

Local adaptation of a native herbivore to a lethal invasive plant.

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Abstract

Understanding the evolutionary processes that influence fitness is critical to predicting species' responses to selection. Interactions among evolutionary processes including gene flow, drift and the strength of selection can lead to either local adaptation or maladaptation especially in heterogeneous landscapes. Populations experiencing novel environments or resources are ideal for understanding the mechanisms underlying adaptation or maladaptation, specifically in locally co-evolved interactions. We used the interaction between a native herbivore that oviposits on a patchily distributed introduced plant that in turn causes significant mortality to the larvae to test for signatures of local adaptation in areas where the two co-occurred. We used whole genome sequencing to explore population structure, patterns of gene flow and signatures of local adaptation. We found signatures of local adaptation in response to the introduced plant in the absence of strong population structure with no genetic differentiation and low genetic variation. Additionally, we found localized allele frequency differences within a single population between habitats with and without the lethal plant, highlighting the effects of strong selection. Our work highlights the potential for adaptation to occur in a fine-grained landscape in the presence of gene flow and low genetic variation.

1 Local adaptation of a native herbivore to a lethal invasive plant.

2 **Running title:** Local adaptation to a novel hostplant.

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16 **Abstract**

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18 species' responses to selection. Interactions among evolutionary processes including gene flow,
19 drift and the strength of selection can lead to either local adaptation or maladaptation especially
20 in heterogenous landscapes. Populations experiencing novel environments or resources are ideal
21 for understanding the mechanisms underlying adaptation or maladaptation, specifically in locally
22 co-evolved interactions. We used the interaction between a native herbivore that oviposits on a
23 patchily distributed introduced plant that in turn causes significant mortality to the larvae to test
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25 sequencing to explore population structure, patterns of gene flow and signatures of local
26 adaptation. We found signatures of local adaptation in response to the introduced plant in the
27 absence of strong population structure with no genetic differentiation and low genetic variation.
28 Additionally, we found localized allele frequency differences within a single population between
29 habitats with and without the lethal plant, highlighting the effects of strong selection. Our work
30 highlights the potential for adaptation to occur in a fine-grained landscape in the presence of gene
31 flow and low genetic variation.

32 **Keywords**

33 evolutionary traps, novel resources, maladaptation, *Pieris*, Brassicaceae, gene flow - selection
34 balance

35 **Introduction**

36 Strong spatially heterogenous selection can lead to locally adapted populations. The
37 ability of populations to adapt to local conditions depends on substantial genetic variation, large
38 effective population sizes and the changes in their habitat (Franch-Gras et al., 2018; Hedrick,
39 2011; Lande, 1976; Lenormand, 2002; Perrier et al., 2020; Simons, 2011). Additionally, local
40 adaptation is maintained through the balance of strong selection and limited gene flow between
41 divergently adapted populations (Blanquart et al., 2013; Hereford, 2009; Savolainen et al., 2013;
42 Whitlock, 2015). Across a heterogeneous landscape, the interplay between gene flow and
43 selection can lead to a geographic mosaic with varying levels of adapted and maladapted sub-
44 populations (Gomulkiewicz et al., 2000; Thompson, 1999a, 2005).

45 Novel environments can derail adaptative evolution by decreasing the fitness of formerly
46 locally adapted individuals (Brady, Bolnick, Angert, et al., 2019; Reed et al., 2003; Robertson &
47 Hutto, 2006; Schlaepfer et al., 2002, 2005). The outcome of interactions between native
48 populations and novel environments depends strongly on population connectivity among
49 divergently adapted populations as well as the population's potential to adapt to novel conditions
50 (Bolnick & Nosil, 2007; Farkas et al., 2015, 2016; Nosil et al., 2019). This often results in
51 persistent mismatches with or maladaptation to the new resources or conditions (Brady, Bolnick,
52 Angert, et al., 2019; Chew, 1977b; Cotto & Ronce, 2014; Singer & Parmesan, 2018; Steward et
53 al., 2022; Steward & Boggs, 2020). Such cases provide unique opportunities for understanding
54 the process of local adaptation because the character, scale and timeline of the environmental
55 change are often well known (Chew, 1977b; Singer & Parmesan, 2018, 2019; Steward et al.,
56 2022; Steward & Boggs, 2020). Identifying and understanding these interactions are critical to

57 understanding species responses to novel environments, especially in the face of anthropogenic
58 disturbance and climate change.

59 Herbivorous insects have a narrow range of hostplants that support development, and
60 those insects form tightly co-evolved interactions with their hosts (Ehrlich & Raven, 1964; Hardy
61 & Otto, 2014; Jaenike, 1990; Joshi & Thompson, 1995; Thompson, 1999b). Non-native or
62 invasive species can break down long-standing interactions and lead to maladaptation in either or
63 both of the interacting partners, either by outcompeting and replacing native species or through
64 eco-evolutionary traps (LaForgia et al., 2020; Richard et al., 2019; Saul & Jeschke, 2015;
65 Schlaepfer et al., 2005; Schweiger et al., 2010). Eco-evolutionary traps arise when previously
66 adaptive traits become maladaptive in the face of novel resources or environments, such that the
67 organisms prefer the low fitness-value novel resource instead of the high fitness-value native
68 resources (Robertson et al., 2013; Robertson & Chalfoun, 2016; Robertson & Hutto, 2006;
69 Schlaepfer et al., 2002).

70 Eco-evolutionary traps resulting from maladaptive oviposition choice have been
71 documented in many insects (Horváth et al., 2007, 2010; Singer & Parmesan, 2018). Lepidoptera,
72 specifically butterflies, are susceptible to laying eggs on low quality novel hosts plants (Nakajima
73 & Boggs, 2015; Singer & Parmesan, 2010, 2019; Steward et al., 2022; Steward & Boggs, 2020;
74 Yoon & Read, 2016). Hence, butterflies serve as an ideal system to test the impacts of novel
75 resources on local adaptation in species interactions. We here use a landscape genetics approach
76 to identify genomic signatures of local adaptation in the specialist herbivore *Pieris*
77 *macdunnoughii* Remington, 1954 (Pieridae; formerly *P. napi macdunnoughii*) (Chew & Watt,
78 2006) in the presence of the toxic invasive plant *Thlaspi arvense* (Brassicaceae). *Pieris*

79 *macdunnoughii* females recognize *T. arvensis* as a potential host plant due to chemical cue
80 similarity with the native brassicaceous hosts and lay eggs on it (Chew, 1977b, 1977a, 1980;
81 Nakajima, 2014; Steward et al., 2019, 2022). However, larvae from eggs laid on *T. arvensis* have
82 significant mortality due to the toxicity of the plant (Chew, 1977b; Steward et al., 2019). We
83 sampled *P. macdunnoughii* in the East River Valley, Gunnison County, Colorado, where the
84 butterfly oviposits on both native host plants and *T. arvensis*.

85 Using genome-wide data from whole genome resequencing of individuals from areas
86 where both the plants occur and from areas where *T. arvensis* is absent, we identified candidate
87 single nucleotide polymorphisms (SNPs) that are under selection in relation to the presence of *T.*
88 *arvensis* in the habitat. We specifically quantify: i) population structure of *P. macdunnoughii* in
89 the East River Valley, ii) gene flow patterns of *P. macdunnoughii* between areas with and without
90 *T. arvensis* and iii) underlying signatures of selection in *P. macdunnoughii* in response to the
91 presence of *P. macdunnoughii* in the landscape. Our work builds on decades of existing research
92 on this system to address the genetic basis underlying maladaptive behavior in the face of novel
93 resources.

94 **Materials and methods**

95 **Study organisms**

96 *Pieris macdunnoughii* is a montane butterfly distributed across the southern Rocky
97 Mountains in North America. *Pieris macdunnoughii* is a specialist herbivore, and females
98 oviposit on and the larvae feed on several native Brassicaceae (mustard) plants including
99 *Boechera* spp., *Cardamine cordifolia*, *Draba aurea* and *Descurainia incana* (Chew, 1977b,
100 1977a, 1980). Pierine larvae have evolved resistance to the toxicity of glucosinolates (secondary
101 metabolites) found in the Brassicaceae by modifying the glucosinolate hydrolysis pathway to

102 form thiocyanates and epithionitriles that are less toxic, instead of forming isothiocyanates
103 (Edger et al., 2015; Wheat et al., 2007). Although pierine butterflies are resistant to glucosinolate
104 toxicity, they are susceptible to species whose glucosinolate concentration or composition differs
105 from the ones with which they have co-evolved locally (Keeler & Chew, 2008).

106 *Thlaspi arvense* (L.) (Brassicaceae) is native to Eurasia and was introduced to the
107 Gunnison basin, Colorado in 1880s (Chew, 1977b). It is an early successional plant and is found
108 in heavily disturbed areas. It has colonized elevations up to 2900m (Best & McIntyre, 1975;
109 Warwick et al., 2011). *Pieris macdunnoughii* females recognize and lay eggs on *T. arvense* as a
110 potential host plant in areas where they both co-occur due to similarity of glucosinolates, and
111 other cues shared among the native and nonnative host plants. However, none of the larvae reared
112 on *T. arvense* in the lab survive, thus leading to an evolutionary trap (Nakajima, 2014; Steward &
113 Boggs, 2020). Previous research has indicated that the chemical profile of *T. arvense* is
114 comprised mostly of the aliphatic glucosinolate sinigrin, whereas native mustards (e.g., *C.*
115 *cordifolia*) that contain both aliphatic and aromatic glucosinolates (Rodman & Chew, 1980).
116 Additionally, recent research has indicated that sinigrin acts as a pre-ingestive deterrent and
117 larvae exhibit delayed feeding on *T. arvense* compared to *C. cordifolia* (Steward et al., 2019).

118 **Population sampling and study area**

119 The study area is situated in the East River valley, Gunnison County, Colorado spanning
120 an altitude from ca. 2800 to 3400m a.s.l. (Figure 1). We sampled 100 female adult butterflies
121 during June - August 2016 from 5 locations and 235 adult butterflies of both sexes during June -
122 August 2019 from 7 locations. Our final sample size consisted of 335 individuals from 11
123 locations. The distance between sampling sites ranged from 1km to 28km. All butterflies were

124 caught using an aerial net. We collected whole body specimens of females and stored them in
125 98% ethanol in 2016. In 2019, 2mm x 2mm square of hindwings and a pair of midlegs were
126 clipped and stored in 98% ethanol, after which the butterflies were marked and released. We also
127 surveyed the presence/absence of *T. arvense* in all our sampling locations. We recorded *T.*
128 *arvense* as present in 2 sites in 2016 (of 5 sites sampled) and 2019 (of 7 sites sampled). Our final
129 sampling list consisted of 3 sites where *T. arvense* was present (of 11 sites sampled).

130 **DNA extraction and WGS library preparation**

131 We extracted genomic DNA from thorax (2016 samples) and from wing clips and a pair
132 of midlegs (2019 samples) using the DNeasy Blood & Tissue Kit (Qiagen, USA). We followed
133 the manufacturer's protocol, with the following modifications: We increased the Proteinase K
134 incubation step to 16 hours, used ice-cold ethanol to precipitate DNA, and heated the elution
135 buffer to 57 °C before DNA elution. We quantified the DNA concentration using a NanoDrop™
136 2000/2000c. We used an in-house Tn5 tagmentation protocol following Andolfatto et al. (2011)
137 for our library preparation. 2-10ng/ul of genomic DNA was tagmented with 1:4 diluted Tn5
138 transposase and 40uM pre-annealed oligonucleotides. The tagmentation was carried out in a final
139 volume of 20uL containing 4uL of 5x TAPS buffer. Samples were incubated at 55 °C for 7
140 minutes followed by rapidly lowering the holding temperature to 10 °C. The Tn5 transposase was
141 deactivated using 2.5uL 0.2% SDS at 55 °C for 7 minutes with a heated lid. PCR-based barcoding
142 and enrichment was carried out in a reaction mixture of 20uL consisting of 2 uL of the
143 tagmentation reaction product, 10uL of 2x OneTaq Hot Start DNA polymerase, 4uL of nuclease
144 free water and 2uL (10uM) each of i7 and i5 Illumina primer. The thermocycling conditions
145 included initial denaturation at 72°C for 3 minutes followed by denaturation at 94°C, followed by

146 10 cycles with 10s at 94°C, 15s at 62°C, 30s at 68°C and final extension for 5min at 68°C. We
147 constructed 5 replicate libraries for each sample, then pooled all the replicates and samples (1920
148 libraries). Size selection of 300-500 bp was carried out using 0.8x volume and 0.6x volume
149 AmpureXP beads. All the libraries were sequenced at MedGenome on an Illumina Novaseq S4
150 platform, generating 150bp paired end reads.

151 **SNP (Single nucleotide polymorphisms) calling**

152 Demultiplexed raw Illumina reads and adapter sequences were trimmed using fastp (Chen
153 et al., 2018) and mapped against the reference genome (Steward et al., 2021) using bwa mem
154 with default parameters (Li, 2013). The resulting SAM files were converted to BAM format,
155 sorted and indexed using samtools (Li et al., 2009). Duplicates were marked and identifier groups
156 were added using PICARD TOOLS with default parameters (“Picard Toolkit,” 2019). The
157 genome dataset had a coverage of $6.84x \pm 2.42x$ (mean \pm s.d) and was quantified using genozip
158 (Lan et al., 2021). Single nucleotide polymorphisms (SNPs) were called across all samples using
159 GATK HaplotypeCaller to generate individual intermediate gVCF files that were then imported
160 using GATK GenomicsDBImport and were finally genotyped using GATK GenotypeGVCFs.
161 We used GATK VariantFilteration to further hard filter SNPs and Indels separately using the best
162 recommended workflow practices (Van der Auwera et al., 2013), and finally used GATK
163 SelectVariants to include only those variants that met the criteria: $QD > 2$, base quality > 30 ,
164 $SOR < 3.0$, $FS < 60$, $MQ > 40$, $MQ \text{ Rank Sum} > 12.5$, $Read \text{ Pos Rank Sum} > 8.0$. We then used
165 VCFtools to retain only high quality bi-allelic variants using the following parameters: minimum
166 allele count = 4, max missing = 0.20, min q= 30, min-mean DP = 6, max mean DP = 60, minDP
167 = 6, max DP = 60. We further used PLINK (Purcell et al., 2007) to prune those loci that were at

168 LD with the following parameters: $r^2 > 0.2$ in a window of 50bp. Our final sample size consisted
169 of 335 individuals and 768,339 SNPs.

170 **Population structure**

171 We used a subset of 735,000 putatively neutral SNPs that were obtained after removing
172 outlier loci (see below) to discern the population structure. Principal Component Analysis (PCA)
173 clustering was carried out using PLINK (Purcell et al., 2007). We complemented our PCA
174 analysis with archetypal analysis following Gimbernat-Mayol et al. (2022), to test biases in the
175 PCA analysis due to irregular sample sizes and to identify latent factors. We removed multi-
176 allelic SNPs prior to running archetypal analysis and performed the analysis with k ranging from
177 2 to 4. We picked the k whose PC1 and PC2 axes together accounted for the most variance. We
178 additionally used ADMIXTURE to corroborate the results from PC and archetypal analysis and
179 to assess population structure and ancestry (Alexander et al., 2009). We performed
180 ADMIXTURE analysis for ancestral clusters K ranging from 1 to 6 and selected the K value with
181 the lowest cross-entropy as the best estimate of population admixture.

182 **Nucleotide summary statistics**

183 We used geoVar (Biddanda et al., 2020) to assess if alleles were shared among
184 habitats/sites or if they were localized to each habitat/site. We estimated allele frequency
185 distribution (site frequency spectrum) in a) all 11 sites in the East River Valley and b) in areas
186 with and without *T. arvensis*. We converted our VCF into a frequency table using geoVar ($n =$
187 768339 SNPs), then calculated the cumulative fraction of variants that contributed to the allele
188 frequency pool for each site/habitat. We used the allele frequency distribution in geoVar in place
189 of widely used SFS (Site Frequency Spectrum) methods (Gutenkunst et al., 2009), since geoVar

190 permits simultaneous comparisons of SFS for more than 2 populations. Furthermore, geoVar
191 allows us to classify minor alleles into common (>5% frequency among all samples), low (1 - 5%
192 frequency), rare (<1% frequency) and unobserved (allele not present).

193 Using VCFtools (Danecek et al., 2011) on our SNP dataset (consisting only of variant
194 sites, n = 768339 sites), we calculated expected and observed heterozygosity of butterflies for
195 sites where *T. arvensis* is present and sites where *T. arvensis* is absent. We used pixy (Korunes &
196 Samuk, 2021) to estimate genome-wide nucleotide (π) diversity. For the input for pixy, we
197 specifically used both the invariant and variant sites in our input VCF file (n = 69339609 sites) as
198 recommended by the authors to prevent bias in our estimates. We used a 10kb sliding window
199 with a 50bp step to calculate π estimates. We used Bartlett's test to test for differences in
200 heterozygosity and nucleotide diversity between the habitats. We used VCFtools on our SNP
201 dataset (consisting only of variant sites, n = 768339 sites) to calculate pairwise F_{st} for all
202 combinations of sites in the East River Valley. We used a sliding window approach with an
203 interval window of 1kb with a 50bp step for estimating pairwise F_{st} .

204 **Migration surface/gene flow**

205 We used divMigrate from the diveRsity package (Keenan et al., 2013) in R (R Core
206 Team, 2020) to estimate directional relative migration rates among sampling sites and
207 PGDSpider (Lischer & Excoffier, 2012) to process the input for divMigrate. The relative
208 migration network was scaled to the largest estimated magnitude based on N_m as a measure of
209 genetic distance. We assessed the significance of the migration network by running 1000
210 bootstrap iterations.

211 We used EEMS (Petkova et al., 2015) to examine spatial variation in migration among
212 populations (demes) and genetic diversity within populations. EEMS estimates genetic
213 differentiation using an isolation-by-distance model of geo-referenced samples to visualize
214 patterns of potential barriers and corridors of gene flow. We used PLINK to convert the VCF file
215 to BED files and calculated a genetic dissimilarity matrix using bed2diff in EEMS. We ran
216 EEMS using 400 and 800 demes and a MCMC run with 1.5 million burn-in iterations followed
217 by 15 million sampling iterations. We ran multiple iterations and adjusted the proposal variance
218 rates after each run until runs converged (Figure S1). The parameters used for the final run are
219 provided in the supplementary file (Table S1). We then combined the final output of all the
220 demes to produce a composite migration and diversity landscape. Migration and diversity rates
221 were illustrated on a log₁₀ scale relative to the overall migration and diversity across the entire
222 landscape, such that a rate of one indicated tenfold higher migration and diversity rates relative to
223 the average.

224 **Genotype-environment association (GEA) analysis**

225 Genome-wide scans for outlier loci were carried out in BayPass (Gautier, 2015). We
226 converted our VCF file into an allele count matrix (n=768339 SNPs) using an in-house python
227 script for BayPass input. We first used the standard covariate model with the Importance
228 Sampling (IS) approximation (-covis) in BayPass to obtain the following parameters: a) the
229 population scaled covariance matrix and b) the XTX (SNP-specific F_{st} that corrects for observed
230 population covariance) score of overall differentiation among sites. We then used the auxiliary
231 covariate model using the MCMC approximation along with the Bayesian auxiliary variable to
232 identify loci that were associated with the presence/absence of *T. arvensis*. We additionally

233 simulated 100,000 loci using the BayPass sim to calibrate the top 1% significance threshold for
234 XTX. For environmental association, we used the Bayes Factor (BF) metric in deciban units (dB)
235 as a measure of support for association with *T. arvensis*. We used a cut-off of $db > 20$ (decisive
236 evidence) in favor of association based on Jefferey's rule (Jeffreys, 1961). Loci associated with *T.*
237 *arvensis* as well as those under adaptive differentiation were used as input in SNPEff (Cingolani
238 et al., 2012) and BlastX (Altschul et al., 1990) to identify the underlying genes and their effects
239 on protein structure.

240 **Results**

241 **Population structure and admixture**

242 Our Principal Component Analysis (PCA) and archetypal analysis using neutral loci
243 failed to separate individuals into distinct clusters and hence did not reveal any population
244 structure among all sites in the East River Valley (PCA: PC1=6.56% and PC2=5.65%; Figure 2a;
245 archetypal analysis: PC1=52.92%, PC2=47.08%; Figure 2b). Our ADMIXTURE analysis
246 corroborated the PCA and archetypal analysis. The lowest Cross Validation (CV) indicated $K = 1$
247 (Table S1). Together, these results indicate that *P. macdunnoughii* sampled from sites throughout
248 the East River valley comprise a single population.

249 **Genetic variation and differentiation**

250 We quantified the relative abundance of private and shared alleles in a) between all sites
251 where *T. arvensis* is absent or present (i.e. between habitats) and b) pairwise comparison among
252 individual sites. The allele frequency distribution for *P. macdunnoughii* in habitats with and
253 without *T. arvensis* were mostly low frequency alleles that were localized to each habitat (Figure
254 3a; 26% of cumulative fraction of variants), followed by 22% of shared and common alleles. The

255 rest of the variants were either rare or low frequency alleles that were localized to each habitat
256 (Figure 3a). In contrast to the habitat-based allele frequency distribution, pairwise comparison of
257 allele frequencies among sites consisted of alleles that were shared/common amongst all the sites
258 (Figure 3b; 14% of the cumulative fraction of variants).

259 Overall, we found higher genetic variation among *P. macdunnoughii* in areas without *T.*
260 *arvense* compared to areas with *T. arvense* (Figure 4a; Bartlett's $K^2=6.95$; $df=1$; $p=0.008$). The
261 observed heterozygosity was less than expected in areas without *T. arvense* (Figure 4a; Bartlett's
262 $K^2=39.63$; $df=1$; $p<0.001$) and the observed vs expected heterozygosity was similar in areas with
263 *T. arvense*. Our genome-wide average estimates of nucleotide diversity (π) diversity indicated a
264 significant increase in π in sites without *T. arvense* compared to sites with *T. arvense* (Figure 4b;
265 Bartlett's $k^2=9.611$; $df=1$; $p=0.001$).

266 Our estimate of genome-wide mean pairwise F_{st} among all combinations of sites was less
267 than <0.001 (Figure S2), indicating low genetic differentiation and high relatedness among sites.
268 A correlation map based on the population co-variance matrix indicated that all sites were weakly
269 positively correlated. Accordingly, the Mantel test did not reveal any signs of isolation by
270 distance ($p=0.14$, $r=0.05$).

271 **Migration**

272 We observed significant variation in the magnitude and direction of relative migration of
273 *P. macdunnoughii* among all sites (Figure 5). The estimated relative migration rates ranged from
274 0 to 1 with an average of 0.56. We observed unidirectional and bidirectional migration from areas
275 with *T. arvense* to areas without *T. arvense* and vice versa. The highest rates of gene flow
276 ($Nm>0.85$) were observed from Gothic Townsite to Elko Park ($Nm=0.89$) and vice versa

277 (Nm=1), and from Quigley Creek to Gothic Townsite (Nm=0.96). Additionally, we observed
278 high rates of relative gene flow to and from Gothic Townsite compared to other sites (Figure 5).
279 Estimated effective migration surface (EEMS) contours revealed the low relative effective gene
280 flow of *P. macdunnoughii* in the East River Valley (Figure 6a). Most of the potential barriers also
281 showed high posterior probabilities in the Bayesian estimation of migration parameters (Figure
282 S3). Posterior probabilities of migration parameters for sites in the Upper East River valley
283 (Copper Creek, Copper Creek 1st Crossing, Gothic Townsite, Rustler's Gulch, Quigley Creek,
284 and Elko Park) were higher (>0.95) compared to the sites in the Lower East River valley (Lower
285 and Upper Brush Creek and Lower and Upper Cement Creek), which had lower posterior
286 probabilities <.90. We also estimated the relative effective genetic diversity of *P. macdunnoughii*
287 in the East River valley. Our analysis highlighted low relative effective genetic diversity of *P.*
288 *macdunnoughii* in all sampled sites (Figure 6b). All sites showed a posterior probability of >.90
289 (Figure S4).

290 **Genotype-environment association (GEA) analysis**

291 The BayPass core model (-covis) allowed us to estimate the scaled covariance matrix of
292 population allele frequencies that quantify genetic relationship among pairs of sites. The results
293 of Ω estimates agreed with our F_{st} estimates indicating that all sites are genetically similar (Figure
294 S2).

295 To identify outlier loci, the XtX (SNP specific F_{st}) estimates were calibrated by analyzing
296 a POD (pseudo-observed data set) of 100,000 SNPs. At the 1% threshold (XtX >26) for POD, we
297 identified 8600 outlier SNPs (Figure S5). Our analysis that included presence/absence of *T.*
298 *arvensis* (as the environmental covariate) under the auxiliary covariate model identified 1008

299 SNPs that were associated with the presence of *T. arvense* (BF >20) (Figure S6). Overall, we
300 identified nine SNPs that were shared by both the XtX outlier loci analysis and environmental
301 association analysis (Figure 7).

302 Our SNPeff analysis indicated that of the nine SNPs, one was identified as a low impact
303 protein coding change in the exon (unlikely to change protein behavior) with the rest of the SNPs
304 impacts classified as modifiers (effects on non-coding regions and/or effects of gene regulation).
305 The BlastX analysis of these regions revealed that these genes were primarily involved in
306 cytoskeletal organization, DNA damage repair, lipidation of chylomicrons in the intestines, eye
307 development, epithelial development, and catalysis of phosphoric acid. Importantly, regions that
308 were under selection and associated with the presence of *T. arvense* have previously been linked
309 to larval development and metabolism (Table 1, S2).

310 **Discussion**

311 *Pieris macdunnoughii* butterflies in the East River Valley comprise a single admixed
312 population. Allele frequency differences between habitats (i.e., presence or absence of *Thlaspi*
313 *arvense*) were largely driven by localized alleles, whereas pairwise allele frequency differences
314 among sites were driven by geographically widespread alleles. Additionally, heterozygosity of
315 butterflies was lower in areas where *T. arvense* occurred compared to areas where the plant was
316 absent. Examining relative and effective gene flow patterns revealed bidirectional, asymmetric
317 relative gene flow and low effective gene flow across the East River valley. Finally, we identified
318 signatures of selection in *P. macdunnoughii* in response to *T. arvense*. Based on our annotation of
319 the assembly, it appears that selection is affecting loci involved in larval ability to feed on *T.*
320 *arvense*, not the adult females' recognition and discrimination of potential hosts. In sum, we

321 identified genetic signatures of local selection in a native herbivore in response to a novel toxic
322 hostplant at a fine geographic scale despite absence of strong population structure and genome-
323 wide differentiation.

324 **Evidence for a single population among sites**

325 Our results did not separate individuals into distinct clusters and indicated that all sites
326 comprise a single population, which accords with previous dispersal estimates for *P.*
327 *macdunnoughii* in this area. Mark-release-recapture surveys in the 1970s and early 2000s
328 estimated *P. macdunnoughii* mean dispersal between 400m and 700m (among dispersants, with
329 approximately 0.42% of recaptures being dispersants) (Nakajima, 2014) which generally exceeds
330 the extent of *T. arvensis* in invaded areas (Nakajima et al., 2013). *P. macdunnoughii* disperses
331 along large elevational gradients (Nakajima, 2014), occurring up to 4500m.a.s.l. Several species
332 of butterflies are known to disperse larger distances, thus increasing gene flow and blurring
333 population boundaries across heterogeneous environments (Kitahara, 2016; Spieth & Cordes,
334 2012; Takami et al., 2004).

335 **Distinct allele frequency distribution patterns between habitats and among sites**

336 Recent research has highlighted the utility of allele or site frequency spectra (AFS/SFS) to
337 identify outlier loci, understand population structure, infer demographic changes, and identify
338 signatures of positive selection. In our study, AFS differences of *P. macdunnoughii* between
339 areas with and without *T. arvensis* was largely driven by alleles that were localized or private to
340 each habitat, even in the absence of strong population structure. However, overall differences
341 among sites in general were largely due to differences in shared, common alleles. This is in
342 contrast with theoretical and empirical work in other systems, which showed that pairwise

343 differences among sites within a population were driven by shared, common alleles (Biddanda et
344 al., 2020; Gutenkunst et al., 2009). For instance, analysis of large-scale human genomic data
345 from the 1000 Genomes Project (1KGP) showed that allele differences between populations were
346 due to localized and rare alleles and differences between pairs of individuals regardless of
347 population origin were due to common variants found globally (Biddanda et al., 2020). Allele
348 frequency differences between populations can result due to local adaptation and/or new
349 mutations specific to the population (Günther & Coop, 2013). In our case, the abundance of
350 localized, low frequency alleles between the habitats could be due to the selection imposed by *T.*
351 *arvensis* leading to putative adaptive alleles.

352 **Mechanism underlying local adaptation with high gene flow and low genetic variation.**

353 Our results highlight important mechanisms that underly local adaptation. Population
354 genetics theory suggests that local adaptation occurs in the absence of gene flow and that high
355 gene flow often leads to homogenization and maladaptation (Bachmann et al., 2020; Farkas et al.,
356 2016; Gandon et al., 1996; Garant et al., 2007; Kirkpatrick & Barton, 1997; Lenormand, 2002).
357 However, recent studies showed that local adaptation occurs in the presence of strong gene flow
358 in certain scenarios (Fitzpatrick et al., 2015; Tigano & Friesen, 2016). For example, in spatially
359 and temporally varying habitats, gene flow augments standing genetic variation and thus
360 increases local adaptation (Blanquart et al., 2012, 2013). Local adaptation can also occur in the
361 presence of gene flow through adaptive introgression and strong selection against the immigrant
362 alleles (Griffiths et al., 2021; Leroy et al., 2020; Rendón-Anaya et al., 2021).

363 Our results indicated that all sites in the East River Valley were panmictic, with high gene
364 flow, low genetic differentiation, and low genetic variation within the population. Nonetheless,

365 we were able to identify strong signatures of local adaptation in *P. macdunnoughii* where *T.*
366 *arvensis* occurred. Eggs laid on *T. arvensis* die before pupation, which introduces fitness costs for
367 immigrant individuals that prefer *T. arvensis* (Nakajima, 2014; Nakajima & Boggs, 2015).
368 Similarly, local adaptation in lava flow lizards (melanism) occurred in the presence of strong
369 gene flow and low genetic variation (Krohn et al., 2019). This is in line with recent theoretical
370 and empirical work that suggests that environmentally driven local adaptation does not lead to
371 genome wide differences or require substantial standing genetic variation unless the underlying
372 traits are linked to reproduction (Feder et al., 2012; Krohn et al., 2019; Shafer & Wolf, 2013).

373 **Selection on oviposition vs larval performance.**

374 We did not find that the fitness costs for naïve females ovipositing on *T. arvensis*
375 translated into signatures of selection at the genomic level. Rather, loci under selection were near
376 or in genes annotated with functions that support larval feeding and performance. Recent studies
377 have found that genomic bases of herbivorous insects' response to novel hostplants are
378 polygenic, involving genes underlying oviposition, larval feeding, larval metabolism, and
379 detoxification (Egan et al., 2015; Gompert et al., 2022; Vertacnik & Linnen, 2017). Similarly, we
380 identified a polygenic response to selection from *T. arvensis*, including 9 loci distributed across
381 five chromosomes, causing 25 variant effects. Identification using SNPeff and BlastX revealed
382 that the majority of the genes were involved in larval development and metabolism and most of
383 the variant changes affected non-coding regions. Non-coding regions in the genome contain
384 regulatory elements that play a critical role in protein assembly, gene expression and regulation
385 and are under purifying selection (Andolfatto, 2005; Bird et al., 2006; Loehlin et al., 2010).

386 Therefore, these variant changes might have significant effects in larval feeding and
387 detoxification.

388 We did not identify any loci under selection that were associated with sensory (olfactory,
389 gustatory, or visual) functions that underlie female oviposition choice (de Fouchier et al., 2017;
390 Engsontia et al., 2014; Ramaswamy et al., 1987; K. Yang et al., 2020). Lepidopteran females use
391 a combination of sensory receptors to identify a potential hostplant (Haverkamp et al., 2018;
392 Renwick & Chew, 1994; Thompson & Pellmyr, 1991). Females use olfaction and visual cues for
393 long range detection of hostplants and the final decision is made after gustatory tactile contact
394 with the hostplant using the first pair of foretarsi in their legs (Ozaki et al., 2011; Ryuda et al.,
395 2013). Differences in oviposition choice in females are driven by underlying differences in their
396 chemosensory repertoire. Thus, our failure to identify chemosensory genes in our outlier analysis
397 suggests that selection might be acting on larvae instead of the females. Existing evidence
398 suggests that error prone oviposition in Lepidoptera females can drive hostplant range expansions
399 and adaptation to new hosts, since repeated oviposition on less suitable/non hostplants imposes
400 selection on the larvae to evolve to feed on the plant (Janz et al., 1994; Nylin et al., 2000; Nylin
401 & Janz, 2009; Stefanescu et al., 2012). Furthermore, *P. macdunnoughii* larvae that survive to
402 later instars can be rescued when they are provided with suitable host plants (*C. cordifolia* or *D.*
403 *incana*) and individual based models have suggested that fine-grained distribution of *T. arvensis*
404 and native host plants can alter population dynamics of the butterfly (Nakajima & Boggs, 2015).
405 Thus, we hypothesize that selection is acting on larvae for longer survival on *T. arvensis*
406 potentially enabling later instar larvae to find suitable native host plants.

407 **Evolutionary constraints on oviposition preference and larval performance**

408 *Pieris macdunnoughii* and its sister taxa in North America are derived from the Eurasian
409 *P. napi* during the last Holarctic speciation event (Chew & Watt, 2006). *Pieris napi* larvae can
410 develop successfully on *T. arvense* and other invasive Eurasian mustards found in the Eastern US
411 including *Alliaria petiolata* (Forsberg, 1987; Friberg & Wiklund, 2019; Prasai & Karlsson,
412 2011). *Pieris oleracea*, another North American species in the *Pieris* species complex has
413 reduced fitness when feeding on *A. petiolata* due to its toxicity to the larvae (Chew, 1977b;
414 Haribal et al., 2001, 2001; Haribal & Renwick, 1998; Huang et al., 1994; Keeler & Chew, 2008).
415 Thus, the North American *Peiris* larvae have lost the ability to develop or have reduced fitness on
416 introduced mustards from Eurasia while females have retained the ancestral preference for
417 hostplants. This mismatch in preference-performance is the underlying cause for maladaptation in
418 the larvae. Our results suggest that any adaptation by *P. macdunnoughii* to be able to use the
419 plant will likely involve evolution of larval ability to develop on *T. arvense* and not of the
420 females' avoidance of oviposition on *T. arvense*. Adaptation in response to *T. arvense* in the
421 larvae would involve longer survival on *T. arvense* followed by rescue in the later instars. Indeed,
422 after decades of maladaptation of *P. oleracea* on *A. petiolata*, the larvae are now able to develop
423 on the plant during its bolting stage but not the rosette stage (Keeler & Chew, 2008). This
424 highlights the importance of understanding the interactions between plant phenology, plant
425 chemical composition and larval performance. Our results suggest that a similar outcome may
426 occur in *P. macdunnoughii* in the event of rapid evolution due to selection. Future research
427 quantifying fine scale spatial and temporal patterns of larval performance coupled with
428 quantifying *T. arvense* distribution and variation in plant chemical profiles might provide insights
429 on the escape from or persistence of the evolutionary traps.

430 **Conclusions:**

431 Instances of interactions between native and non-native species are projected to increase
432 across the globe due to range shifts, competition from invasive species and climate change.
433 Understanding the effects of these interactions requires careful dissection of the ecological and
434 molecular processes that mediate these interactions. Our results build upon decades of research
435 aimed at understanding the causes and maintenance of evolutionary traps, to elucidate the
436 molecular response and potential for adaptation to novel resources (Brady, Bolnick, Barrett, et
437 al., 2019; Farkas et al., 2015; Gilroy & Sutherland, 2007; Robertson et al., 2013; Schlaepfer et al.,
438 2002, 2005). As insects are currently experiencing a global decline (Hallmann et al., 2017;
439 Nakajima & Boggs, 2015; Wagner et al., 2021), the *Pieris macdunnoughii* - *Thlaspi arvense*
440 system can serve as a model to understand and predict the outcomes of these interactions in
441 insects even at a fine geographical scale. This work identifies the molecular mechanisms that
442 underlies the evolutionary trap, thus pioneering *Pieris macdunnoughii* - *Thlaspi arvense* as a
443 model system to understand maladaptation and evolutionary traps in the face of climate change
444 (Chew, 1977b; Nakajima, 2014; Steward et al., 2019, 2021, 2022; Steward & Boggs, 2020). Our
445 work also highlights the potential of adaptation to occur in a fine-grained landscape in the
446 absence of genetic variation and high gene flow.

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888

889 **Data accessibility statement**

890 Raw sequence reads will be deposited in the NCBI SRA database upon acceptance and prior to
891 publication. Metadata will also be stored in the NCBI SRA using the Invertebrate MIxS version
892 1.0 Package. The source code for data analysis will be made available on Github upon acceptance
893 and prior to publication.

894 **Benefits generated**

895 Benefits from this research will ensue from the data and results shared from this study on public
896 databases as described above.

897 **Author contributions**

898 NR and CLB designed the research. RAS collected specimens in 2016; NR collected specimens
899 in 2019. NR did the molecular work and analyzed the data. NR wrote the draft manuscript. All
900 authors contributed to manuscript revisions. CLB provided oversight for all stages of the work.

901 **Tables**

902 Table 1: Annotation of outlier loci identified by BayPass XtX ($XtX > 26$) and BF($dB > 20$),
903 associated isoforms and predicted effects. Annotation was carried out using SNPeff and BlastX.

Loci (SNP) position	Chromosome	Codon affected	Intervals affected	BlastX identification	Annotation
156011	1	STOP	153581 - 153583	cilia- and flagella-associated protein 410 isoform	Regulation of cell morphology and cytoskeletal organization (Bai et al., 2011); Involved in DNA damage repair (Lai et al., 2011).
3430127	7	STOP	3428853 - 3428855	uncharacterized protein	
11523752	9	START	11545167 - 11545169	ARFRP 1	Lipidation of chylomicrons in the intestine and required for VLDL lipidation in the liver (Jaschke et al., 2012).
4595201	9	START	4592359 - 4592361	homeobox protein Hox-A3-like	Part of a developmental regulatory system that provides cells with specific positional identities on the anterior-posterior axis (Gaudet et al., 2011).
4595201	9	STOP	4595024 - 4595026	homeobox protein Hox-A3-like	Part of a developmental regulatory system that provides cells with specific positional identities on the anterior-posterior axis (Gaudet et al., 2011).
4595201	9	START	4597927 - 4597929	retinal homeobox protein Rax-like	Plays a critical role in eye formation by regulating the initial specification of retinal cells and/or their subsequent proliferation (Kimura et al., 2000).
4595201	9	STOP	4599618 - 4599620	retinal homeobox protein Rax-like	Plays a critical role in eye formation by regulating the initial specification of retinal cells and/or their subsequent proliferation (Kimura et al., 2000).
11703864	14	START	11697968 - 11697970	DOCK 4	Functions as a guanine nucleotide exchange factor (GEF) that promotes the exchange of GDP to GTP, converting inactive GDP-

					bound small GTPases into their active GTP-bound form (Yan et al., 2006).
11703864	14	STOP	11701979 - 11701981	inx3	Structural components of the gap junctions. Essential for proper epithelial development of the epidermis (Lehmann et al., 2006).
5741973	14	START	5730927 - 5730929	mediator of RNA polymerase II transcription subunit 15	Required for activated transcription of the MtnA, MtnB and MtnD genes. Negatively regulates sex comb development (Boube et al., 2000); Required for cholesterol-dependent gene regulation. Positively regulates the Nodal signaling pathway (F. Yang et al., 2006).
5741973	14	STOP	5742219 - 5742221	tripartite motif-containing protein 45	May act as a transcriptional repressor in mitogen-activated protein kinase signaling pathway (Wang et al., 2004).
261573	xfSc0000009	STOP	261794 - 261796	protein ALP1-like	Encodes an alkaline phosphatase. Alkaline phosphatases catalyze the hydrolysis of monoesters of phosphoric acid and a transphosphorylation reaction in the presence of large concentrations of phosphate acceptors (Harper & Armstrong, 1972).
261573	xfSc0000009	STOP	263610 - 263612	Transposase	
577356	xfSc0000009	START	573908 - 573910	uncharacterized protein	

905 **Figure legends**

906 **Figure 1:** Map of sites where butterflies were collected in the East River valley. Sites in red
907 represent areas where *Thlaspi arvense* does not occur and sites in brown represent areas where *T.*
908 *arvense* is present. The size of the circle represents the magnitude of sample sizes from each
909 location.

910 **Figure 2: a)** Genome-wide Principal Component Analysis (axes 1 and 2) using neutral loci
911 indicating a mixed population without distinct clustering by site. **b)** Archetypal analysis (axes 1
912 and 2) with $k=3$ again showing a mixed population without site-specific clusters.

913 **Figure 3: a)** Relative abundance of allele frequency variants of *P. macdunnoughii* in areas with
914 *T. arvense* and areas without *T. arvense*. The codes depicted in the figure represents the
915 frequency of the minor allele in the habitat (U-undetected: No alleles present, R-rare: <1% minor
916 allele frequency (MAF), L-low frequency: 1%-5% MAF, C-common: >5% MAF). The
917 percentages and their corresponding numbers indicate the number of variants in each class and
918 are grouped based on SNP identity. Grey rows represent alleles whose contribution was not
919 significant to the differences between comparisons. **b)** Relative abundance of variants of *P.*
920 *macdunnoughii* in our study area. Ucc: Upper Cement Creek, Ubc: Upper Brush Creek, Cc:
921 Copper Creek, 1c: 1st Crossing, 401: 401 site, Rg: Rustler's Gulch, Qc: Quigley Creek, Ep: Elko
922 Park, Lcc: Lower Cement Creek, Lbc: Lower Brush Creek, Gt: Gothic townsite.

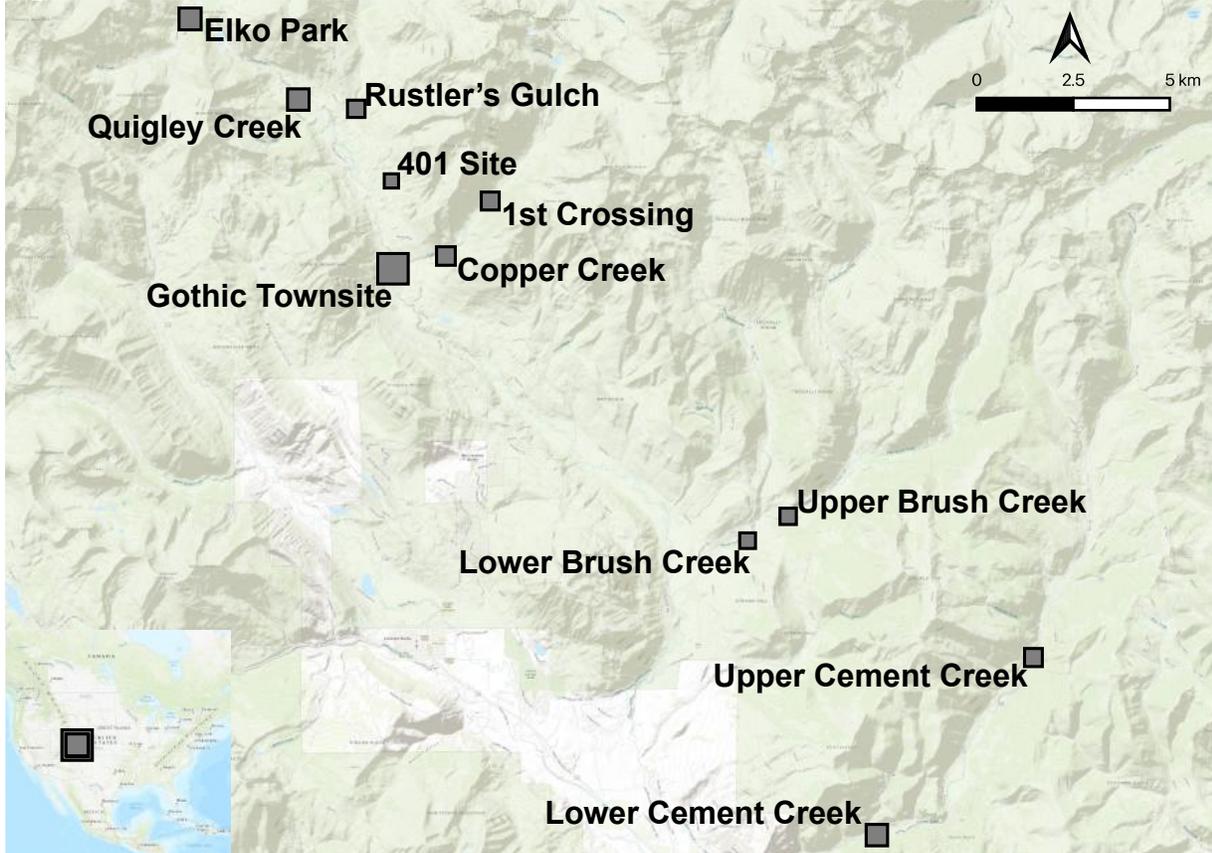
923 **Figure 4: a)** Genome-wide variance in expected and observed heterozygosity of *P.*
924 *macdunnoughii* in the East River valley. Bartlett's tests were used to compare observed
925 heterozygosity between habitats where *T. arvense* was present or absent, and observed versus
926 expected heterozygosity within each of these habitat types. **b)** Genome-wide variance in
927 nucleotide diversity (π) of *P. macdunnoughii* in habitats without and with *T. arvense* (Bartlett's
928 test, $***P < 0.001$). .

929 **Figure 5:** Relative migration rates of *P. macdunnoughii* among the 11 sites in the East River
930 valley estimated using divMigrate based on the number of migrants per generation (Nm)). Colors
931 indicate migration levels (low = blue, high = red), and arrows indicate direction of migration.
932 Solid lines represent unidirectional migration and dashed lines represent bi-directional migration.
933 Colors in bidirectional migration correspond to the migration level of the closest arrow.
934 Migration routes are only shown for $Nm > 0.55$ and migration routes from areas with *T. arvense*
935 to areas without *T. arvense* and vice-versa.

936 **Figure 6: a)** Estimated Effective Migration Surfaces and **b)** Estimated Effective Diversity
937 Surfaces for *P. macdunnoughii* in the East River Valley. The migration and diversity rates,
938 $\log(m)$ and $\log(q)$ represent gene flow or genetic diversity barriers and corridors in the habitat,
939 respectively. Each value corresponds to a 10-fold increase (blue) or decrease (orange/brown) in
940 migration or genetic diversity compared to the null hypothesis of isolation by distance (white).

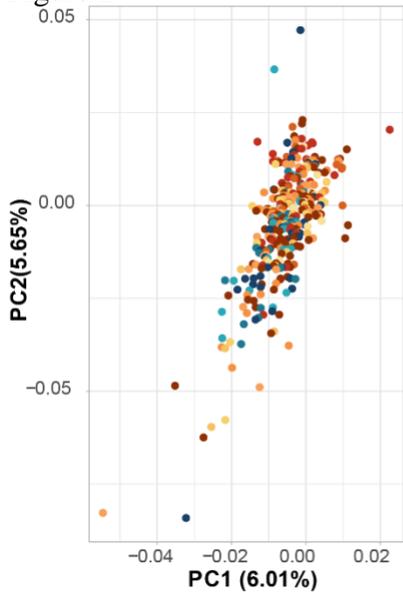
941 **Figure 7:** Pairwise comparison of median SNP XtX and BFmc values from three independent
942 BayPass analyses for adaptive differentiation associated with *T. arvense* presence in the habitat.
943 The vertical dashed line represents the 1% POD (Pseudo Observed Dataset) significance
944 threshold (XtX=26), and the horizontal dashed line represents the 20-dB threshold for BFmc.
945 Blue dots represent the 9 loci of interest (outliers for both XtX and BFmc).

946 Figure 1

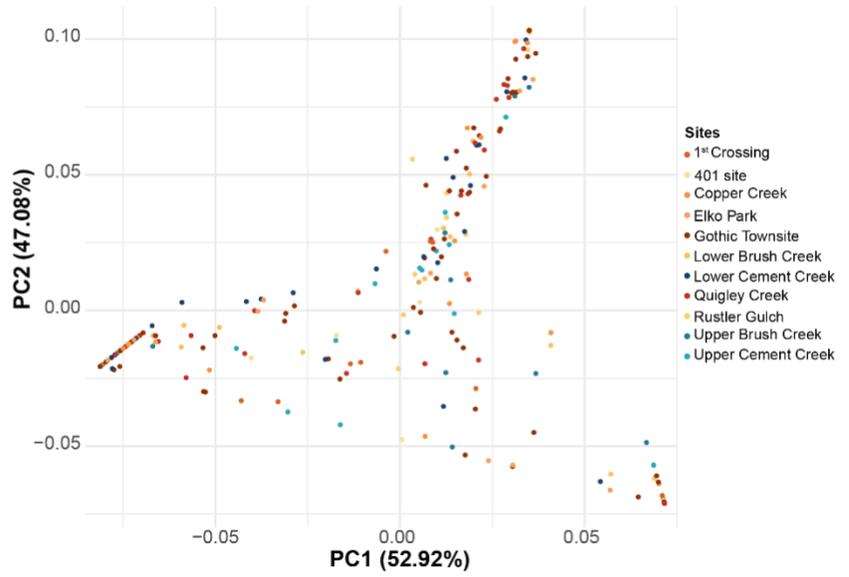


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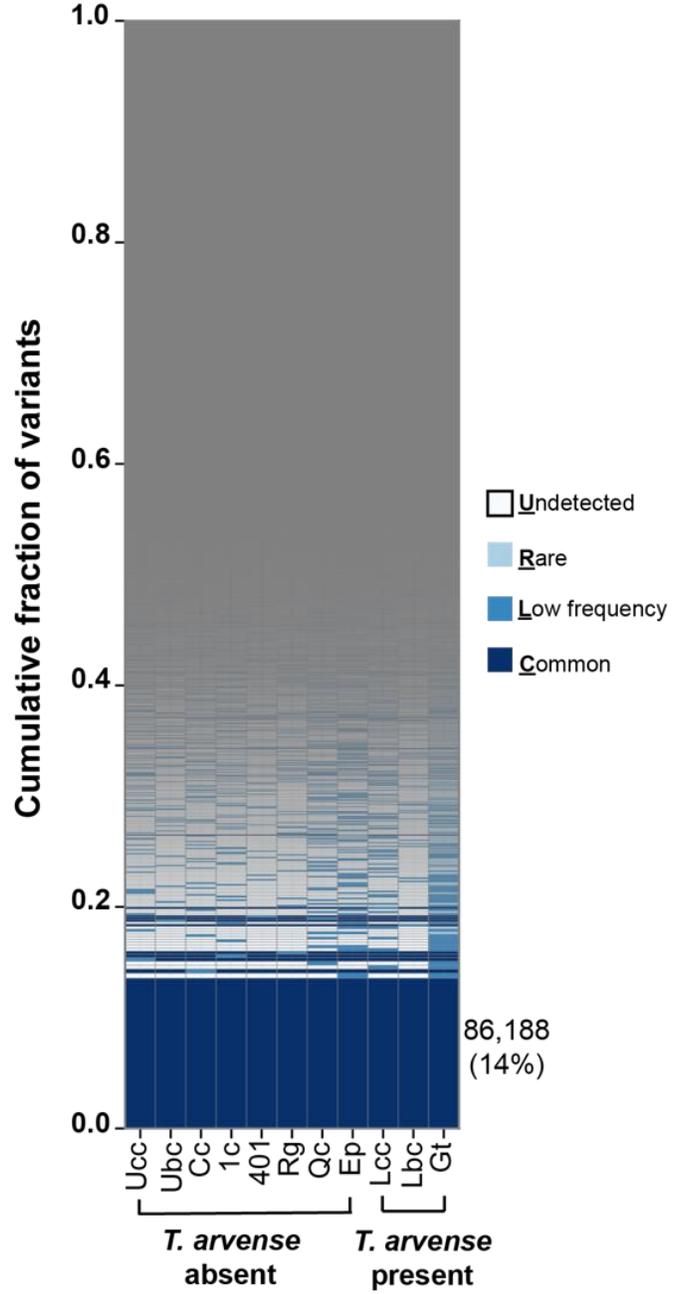
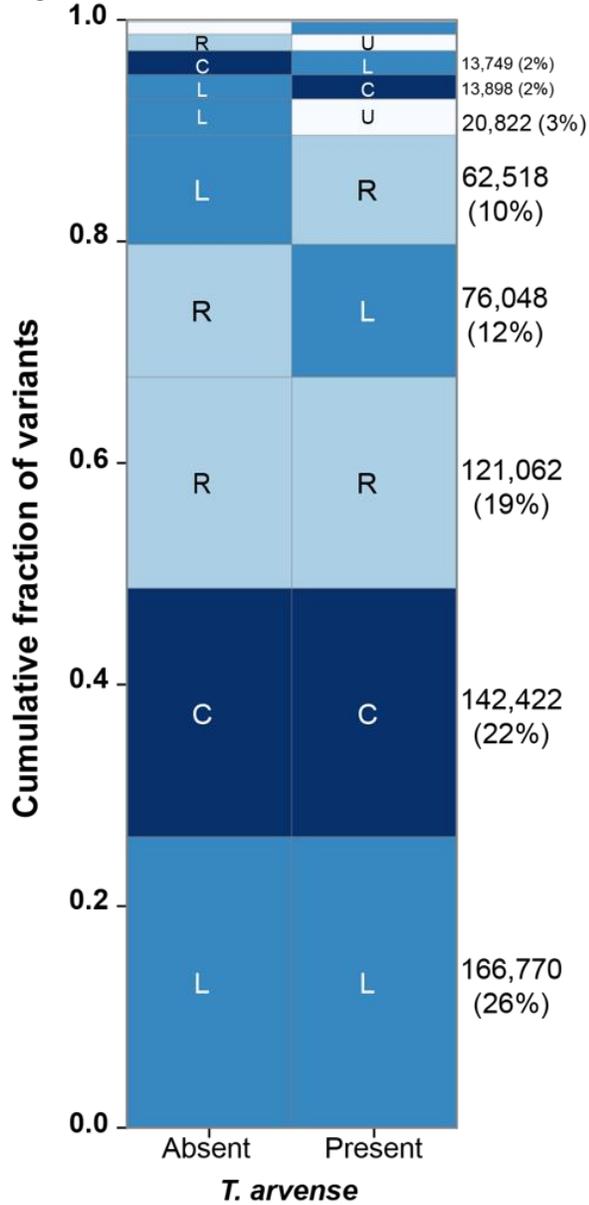
948 Figure 2



949

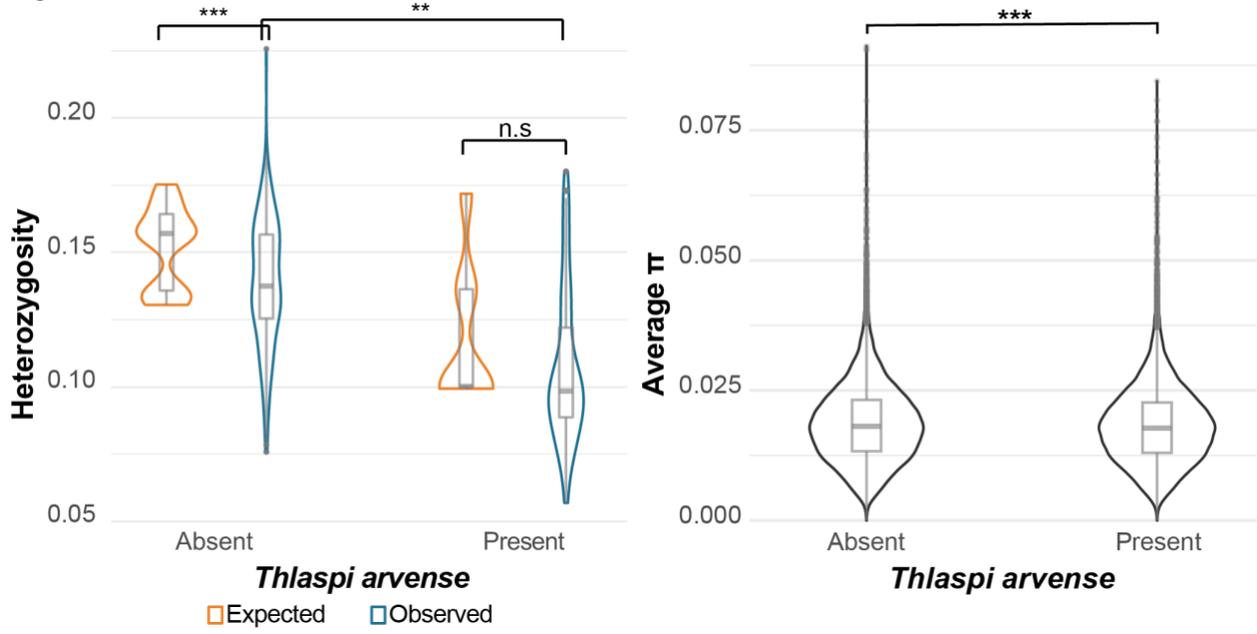


950 Figure 3



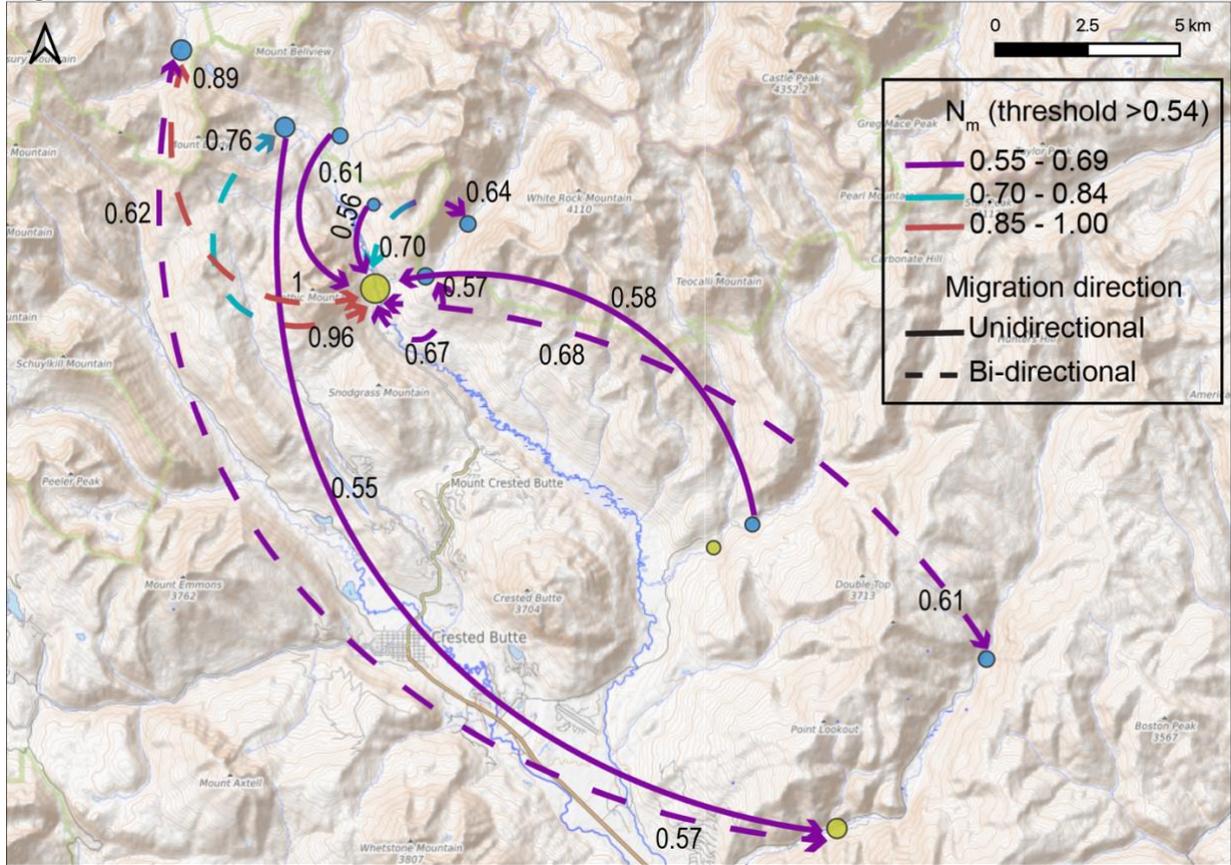
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952 Figure 4

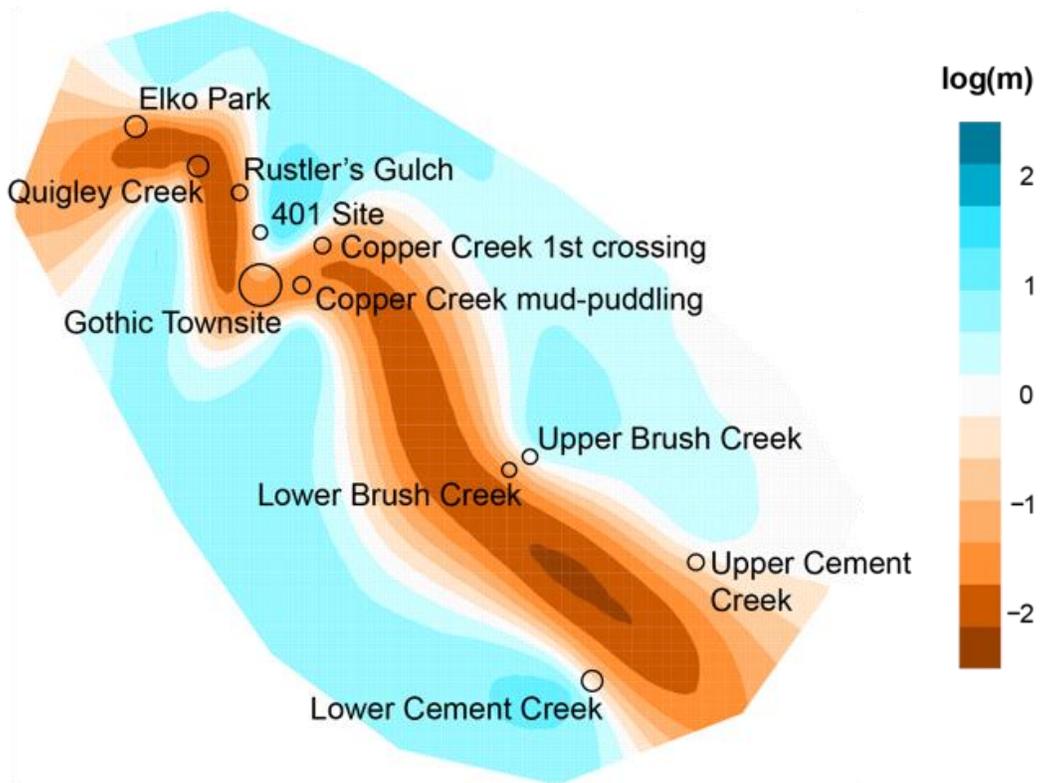


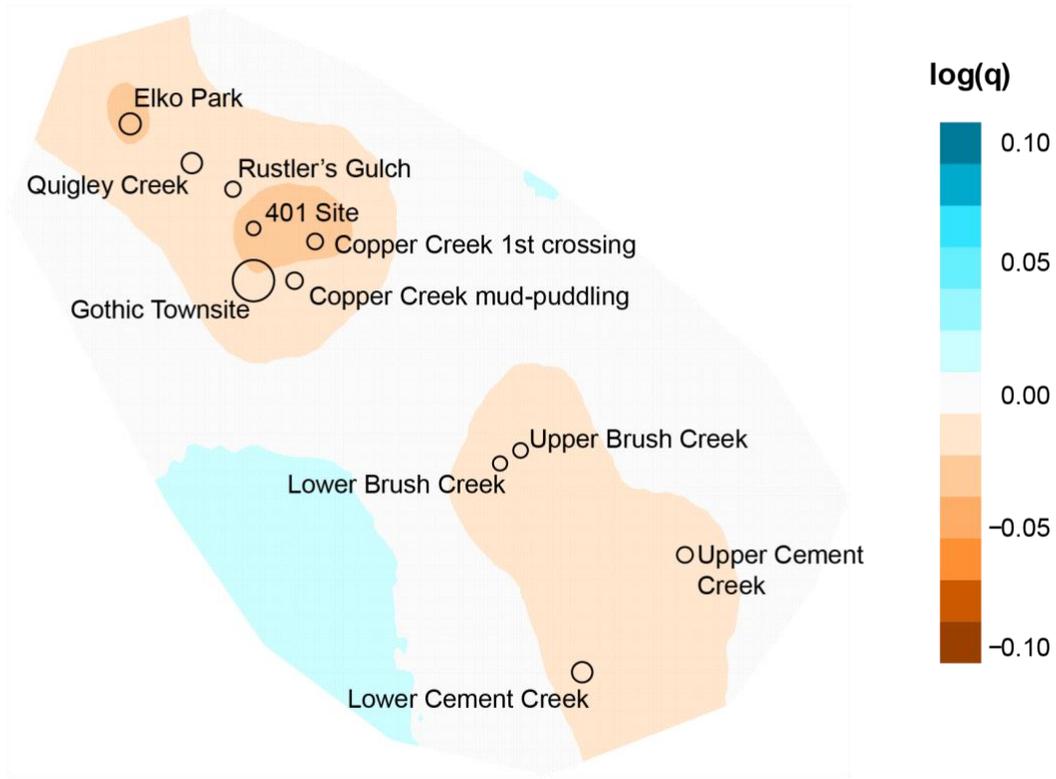
953

954 Figure 5

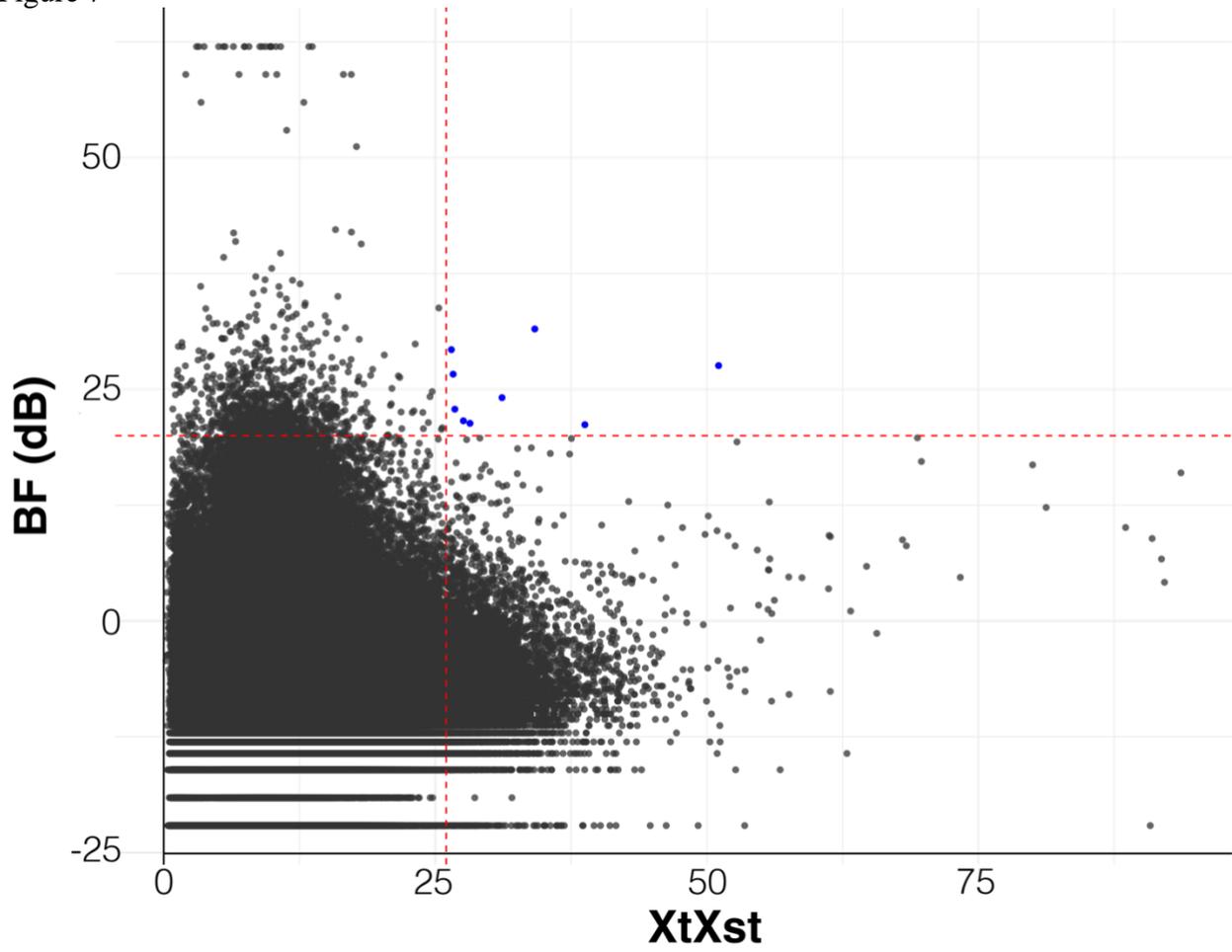


955





960 Figure 7



961