Contemporary hybridization among *Arabis* floodplain species creates opportunities for adaptation

Authors:

Neda Rahnamae[1,2], Lukas Metzger[3,4], Lea Hördemann[1], Kevin Korfmann[5], Abdul Saboor Khan[1], Craig I. Dent[6,7], Samija Amar[6], Raúl Y. Wijfjes[8], Tahir Ali[1], Gregor Schmitz[1], Benjamin Stich[9], Aurelien Tellier[3], Juliette de Meaux[1,*]

[1]Institute for Plant Sciences, Biocenter, Cologne, Germany

[2]Institute of Plant Ecology and Evolution, Heinrich-Heine University Düsseldorf, Düsseldorf, Germany

[3]Professorship for Population Genetics, Department of Life Science Systems, Technical University of Munich, Freising, Germany

[4]Research Department for Limnology, Mondsee, University of Innsbruck, Innsbruck, Austria

[5]Institute of Ecology and Evolution, University of Oregon, Eugene, OR, USA

[6]Department of Chromosome Biology, Max Planck Institute for Plant Breeding Research, Cologne, Germany

[7]CEPLAS: Cluster of Excellence on Plant Sciences, Heinrich-Heine University, Düsseldorf, Germany

[8]Faculty of Biology, LMU Munich, Planegg-Martinsried, Germany

[9]Julius Kühn Institute (JKI), Federal Research Centre for Cultivated Plants, Institute for Breeding Research on Agricultural Crops, Sanitz, Germany

*For correspondence (e-mail: jdemeaux@uni-koeln.de).

Abstract

Hybridization between closely related species is increasingly recognized as a major source of biodiversity. Yet, whether it can create advantageous trait combinations while purging harmful alleles remains unknown. We studied *Arabis nemorensis* and *A. sagittata*, two endangered species that currently hybridize in a single hotspot. We measured 22 phenotypic traits and mapped their genetic basis in an F2 population, after generating high quality genome

assemblies for both species. In total, 58 QTLs were identified for 20 traits, with additive and dominance effects best fitting Gaussian and logistic distributions, respectively. Six large-effect QTLs were linked to significant hybrid fitness loss. Two genomic regions showed strong transmission bias favoring *A. sagittata* alleles, potentially accelerating their introgression. However, 48% of QTLs were unlinked to reduced fitness or segregation distortion and may generate genotypes exceeding parental performance. Notably, a major QTL affecting flowering time explained 23% of phenotypic variation and implicated *TFL1* as a candidate gene for life history adaptation. While most QTLs lacked overlap with past selective sweeps, indicating limited recent positive selection, 5 of 7 QTLs for rosette size overlapped with sweep signatures in the parental lineages. Overall, our findings offer unique insights into incipient stages of hybridization.

Keywords: hybridization, QTL mapping, meiotic drive, heterozygote advantage, competitive grasslands, *Arabis* genome.

Introduction

Biodiversity is crucial for the structure and functioning of ecosystems (Staudinger *et al.*, 2012). The direct impact of human activity and the escalating threat of climate change have initiated the sixth Mass Extinction (Cochrane *et al.*, 2016; Ceballos *et al.*, 2020; Cowie *et al.*, 2022; IPCC, 2023), sparking global concern (Hooper *et al.*, 2002; Eichenberg *et al.*, 2021; Theissinger *et al.*, 2023) and prompting investigation into how species will cope (Moore & Hendry, 2009; Bontrager & Angert, 2019; Schlaepfer & Lawler, 2023). For species endangered by rapid shifts in the environment, the acquisition of pre-adapted alleles via hybridization could provide plants with a strategy to immediately combat climate change (Hansen *et al.*, 2012; Anderson *et al.*, 2012; Chunco, 2014; Brauer *et al.*, 2023).

Hybridization, the interbreeding of individuals from genetically distinct populations or species, occurs frequently among close relatives (Blanckaert *et al.*, 2023; Peñalba *et al.*, 2024; Rosser *et al.*, 2024). It can have both advantageous and detrimental consequences on the species receiving gene flow (Peñalba *et al.*, 2024). First, interspecific hybridization can enable locally adapted alleles across species barriers to be transferred and thus enhance the adaptive potential of species (Seehausen, 2004; Pfennig *et al.*, 2016; Abbott, 2017). Indeed, hybridization between populations with different ecological specializations can give rise to new, viable, and fertile hybrids equipped with novel trait combinations. Such combinations may improve the fitness of the population or even enable previously untapped habitats to be colonized (Buerkle *et al.*, 2000; Rieseberg *et al.*, 2003; Mallet, 2007; Abbott *et al.*, 2013; Blanckaert *et al.*, 2023). Because this process can contribute to the rescue of endangered

species, it may help maintain diversity, as seen in genus *Pachyclasdon*, which appears to
have survived the Last Glacial Maximum thanks to genetic information transferred through
hybridization in alpine refugia on New Zealand's South Island (Becker *et al.*, 2013). Early-
flowering alleles of *Helianthus argophyllus* were enriched among locally adapted introgressed
genomic fragments in *Helianthus* spp and *Betula platyphtylla* (Todesco *et al.*, 2020; Nocchi *et*
al., 2023). Hybrid populations between generalists and narrow range endemic rainbowfishes
(*Melanotaenia* spp.) were well adapted to environmental changes thanks to the effects of
introgression (Brauer *et al.*, 2023). Hybridization has even been proposed to promote the
formation of new species, a phenomenon referred to as "hybrid speciation"; however, whether
a new species actually results is controversial (Anderson & Stebbins, 1954; Ellstrand &
Schierenbeck, 2000; Schumer *et al.*, 2015).

The detrimental effects of hybridization are often more readily detected than its advantageous
effects. Indeed, allelic incompatibilities can cause a massive fitness breakdown, when gene
pools reunite after long periods of evolution in isolation (e.g., *Drosophila* [Masly & Presgraves,
2007; Cooper *et al.*, 2018]; *Mimulus* [Zuellig & Sweigart, 2018]; *Mus* [White *et al.*, 2011; White
et al., 2012; Wang *et al.*, 2015]; *Xiphophorus* [Schumer *et al.*, 2014; Schumer & Brandvain,
2016; Powell *et al.*, 2020; Moran *et al.*, 2024]; Bomblies & Weigel, 2007; Presgraves, 2010;
Coughlan & Matute, 2020; Li *et al.*, 2022). The overall fitness of the hybridizing population will
be reduced if too many resources are used to produce poorly performing hybrids (Rhymer &
Simberloff, 1996; Todesco *et al.*, 2016; Goulet *et al.*, 2017). This phenomenon, sometimes
described as "demographic swamping," elevates the risk of extinction but may also select for
allelic variations that reinforce species isolation (Hopkins, 2013; Goulet *et al.*, 2017; Ma *et al.*,
2019; Brauer *et al.*, 2023).

Despite the interest in the positive consequences of hybridization and the abundant evidence
for allelic incompatibilities (Bomblies & Weigel, 2007), we know little about how the positive
and negative effects of hybridization interact, much less about how a genotype carrying
adaptive alleles might emerge – especially against a background in which detrimental effects
have been recombined out. Such "super genotypes," although rare in the offspring of the first
generation of hybrids (many of which may perform poorly), may represent new and exceptional
combinations of adaptive alleles that in selfing species can determine the evolutionary success
of hybridization. In addition, much focus has been placed on showing that introgressed alleles
bring adaptive advantages to the recipient species (Edelman and Mallet 2021). Yet, the role
that natural selection has played in shaping the introgressing alleles of the donor species has
seldom been examined. Addressing these two challenges is best achieved when hybridization

can be studied as it proceeds, as this allows direct observation of the recombination and
selection dynamics shaping introgressed variation.

Here, we focused on a hybridization hotspot located along the banks of the Rhine River near
Mainz, Germany, where two endangered Arabis species hybridize in a floodplain meadow
(Dittberner *et al.*, 2019; Dittberner *et al.*, 2022). *Arabis nemorensis*, a species within the
Brassicaceae family (*Arabis hirsuta* tribe), inhabits floodplain meadows and is currently in a
critical state in Central Europe, requiring special attention from conservation authorities
(Schnittler & Günther, 1999; Burmeier *et al.*, 2011). *A. nemorensis* is self-pollinating and
exhibits low levels of nucleotide diversity. Its endangered status is further intensified by its
unique ecological requirements, and the loss of its natural habitat (Hölzel, 2005; Burmeier *et*
al., 2011; Mathar *et al.*, 2015; Dittberner *et al.*, 2019). *A. sagittata*, another member of the
same phylogenetic tribe (Karl & Koch, 2014), is morphologically very similar but commonly
found in calcareous grasslands and thus possibly more tolerant to drought (Dittberner *et al.*,
2022). *A. sagittata* was recently observed in floodplains, where it naturally hybridizes with *A.*
nemorensis (Dittberner *et al.*, 2019; Dittberner *et al.*, 2022).

Introgression analysis has revealed that gene flow between *A. nemorensis* and *A. sagittata*
has happened in the past, but contemporary hybridization is restricted to the sympatric
population, where intraspecific genetic variation is extremely low (Dittberner *et al.*, 2019;
Dittberner *et al.*, 2022). Additionally, population genetics analyses estimated the divergence
between *A. nemorensis* and *A. sagittata* occurred around 900,000 generations ago, with
practically complete isolation after the last glaciation (Dittberner *et al.*, 2022).

Using the F2 progeny of a cross between two representative genotypes of *A. nemorensis* and
A. sagittata from the hybridizing hotspot, we addressed the following questions: (1) What are
the phenotypic differences of relevant ecological traits? (2) What is the genetic architecture of
interspecific differences in this hybridizing hotspot? (3) Can a new combination of ecologically
relevant traits arise and establish after hybridization? (4) Do QTL regions enriched in genomic
regions carry footprints of past selection? Our study confirmed the extent of the genetic
differences underpinning phenotypic divergence between two species. In addition, we found
that some F2 hybrids show extreme trait values, despite their generally lower seed production
compared to parental lines. Interestingly, incompatibility QTLs have a simple genetic basis,
and some ecologically relevant QTLs are independent from the incompatibility QTLs.
Collectively, these results indicate that a few offspring of hybridized species may potentially
harbor a genotype with a combination of properties that none of the parents have.

Materials and Methods

Common Garden Experiment and Phenotyping

We crossed sympatric *A. nemorensis* genotype 10 with the *A. sagittata,* genotype 69, collected from the banks of the Rhine River near Mainz in Riedstadt, Hessen, Germany in 2015 and fully sequenced (Dittberner *et al.*, 2019; Dittberner *et al.*, 2022). Because nucleotide diversity within the population is very low (*A. nemorensis* π = 4.37e-5 and *A. sagittata* π = 1.32e-5, synonymous sites, Dittberner *et al.*, 2022), we assume here that differences between these genotypes will predominantly reflect differences between species. Plants were reciprocally crossed to generate F1s, and we noted that their fitness was comparable to that of the parental species (Fig. S1). Seedlings were grown in the greenhouse of the Experimental Garden of the University of Cologne, and the seeds of the first generation of selfing (F2) were harvested. We sowed F2 seeds in trays on 05.10.2019, and after two weeks of vernalization and germination we transplanted seedlings into 7x7cm pots filled with *Topferde* soil (Einheitserde, Sinntal-Altengronau, Germany), placing one seedling per pot. In total, 1,204 individual plants (both hybrids and parental replicates) were distributed across 43 trays, which were put in cold frames to accelerate growth and prevent frost damage from 08.11 to 13.12.2018. On 13.12.2018, we transferred trays to bird-protected cages under seminatural conditions. On 19.11.2018, photographs of each tray were taken using a Canon EOS and rosette sizes were measured using ImageJ. Also, on 19.12.2018, we started harvesting leaves from plants that were big enough (~50 mg of leaves per plant). Harvesting continued in January, February, and March 2019. Harvested rosette leaves were stored at -80°C for DNA extraction.

On 01.03.2019, eight trays were moved to cold frames for a 10-day period to recover from potential stress. A total of 199 F2 individual plants, along with seven *A. nemorensis* and seven *A. sagittata*, were submerged in transparent boxes, each containing 17 L of water for seven weeks. Plants were randomized and distributed across eight boxes, which were kept in cold frames throughout the submergence experiment. After seven weeks of complete submergence, plants were removed from water and left to recover. Then, their survival status was documented. Following a two-week recovery period, we again recorded the survival status of plants, categorizing plants as either dead, survived with new leaf growth, or survived and bolted. During recovery, pots remained in cold frames and were watered as necessary.

Between November 2018 and April 2019, we recorded more than 20 phenotypic traits for each plant. These traits were grouped into five categories: (1) fitness: seed mass from 10 randomly selected siliques or fertility score and survival status after submergence; (2) growth: rosette diameter at four time points (19.11.2018, 08.01.2019, 07.02.2019, 13.03.2019), inflorescence

165   height, and number of side and ground shoots; (3) timing: bolting and flowering times; (4)
166   rosette leaf Ttraits: stem leaf number, leaf length, petiole length, lamina length, lamina
167   length/width ratio, leaf margin type (serrated or smooth), leaf width, and petal length; and (5)
168   stem traits: stem leaf density, stem leaf length and width, and stem height.

Phenotypic Analyses

To assess phenotypic differences between the two species, we used a generalized linear
model (trait ~ Tray, with quasipoisson distribution of error) to analyze each trait, controlling for
experimental effects, in R. A t-test from the stats package in R (version 3.6.2) was then applied
to the residuals of the models to calculate the significance of differences (p-value < 0.05)
between the two groups (*A. nemorensis* and *A. sagittata*, each with 35 replicates) for each
trait. We used ggplot2 (version 3.5.1; Wickham, 2011) to visualize the distribution of
phenotypes for both F2 and parental replicates in a single plot per trait, and to inform our
understanding of transgressive segregation within the F2 population (Suppl. File 1).

For genetically based phenotypic correlation between individuals of the F2 population, we
employed a mixed-effects model approach (trait ~ Cross/TrayBlock) that accounted for
random variation attributable to experimental blocks and trays, and cross direction due to
reciprocal cross. Residuals were extracted for each trait, and pairwise Spearman correlations
were calculated using the *corr.test* function from the psych package in R (version 2.4.6.26);
pairwise deletions were applied to handle missing data. Correlation significance was assessed
at α = 0.05. The resulting correlation and p-value matrices were used to construct a network
in which only significant correlations were retained. A network graph was generated using the
igraph (version 2.1.1) and ggraph (version 2.1.1) packages. Each node represents a
phenotypic trait, and edges represent statistically significant pairwise correlations. Edge color
indicates the sign of the correlation (green for positive, brown for negative), and edge
thickness is scaled to the absolute strength of the correlation. Curved edges and partial
transparency were used to enhance visual clarity, and node labels were plotted within stylized
trait circles. The *ggcorrplot* function from the ggcorrplot package (version 0.1.4.1) was used to
visualize the heatmap.

DNA Extraction and RAD-seq Library Construction

DNA was extracted from leaves stored at -80°C using the Nucleospin® 8 Plant II protocol.
Using the restriction-site-associated DNA sequencing (RAD-seq) protocol described in
Dittberner *et al.* (2019), we analyzed the genomes of 801 F2 individuals. DNA was quantified
with Qubit® 3.0. Fluorometer. For each sample, a total of 250 ng of genomic DNA was

digested with the KpnI-HF restriction enzyme (New England Biolabs). Each digested sample

was then ligated with one individually barcoded modified Illumina P1 adapter containing 5 bp

random nucleotides to PCR duplicates to be removed in the steps that follow. Twenty different

barcodes were used, and 52 pools of 20 barcoded individuals each were constructed, with all

pools standardized to equal volume and concentration. Libraries were sequenced in four

separate runs at the Cologne Center for Genomics (CCG) on the Illumina NovaSeq platform.

For 46 pools, sequencing length was 2x100 bp, generating approximately 3 million reads per

individual; for the remaining 6 pools, sequencing length was 2x150 bp, generating

approximately 6 million reads per individual. These RADseq libraries covered, on average,

approximately 2% of the 248 Mb genome.

Genome Assembly and Annotation

In order to improve previous genome assemblies, the DNA of both parental lines was

sequenced with PacBio HiFi technology and Hi-C data. We used Jellyfish (version 2.3.0;

Marçais & Kingsford, 2011) to count k-mers of size 21 in the 2.18 and 1.98 million reads

obtained for *A. nemorensis* and *A. sagittata*, respectively. The k-mer histogram generated by

Jellyfish was then processed with GenomeScope (version 2.0; Ranallo-Benavidez *et al.*, 2020)

to estimate genome size, heterozygosity, and repetitiveness. HiFiAdapterFilt (version 2.0.0;

Sim *et al.*, 2022) was used to remove residue PacBio adapter sequences from the HiFi reads.

Then, hifiasm was used again to assemble the filtered HiFi reads (version 0.16.1; Cheng *et*

al., 2021; Cheng *et al.*, 2022) with integration of Hi-C data.

Hi-C reads were aligned to the draft contigs using BWA (version 0.7.17; Li & Durbin, 2009),

and the resulting alignments were processed with the Juicer pipeline (version 1.6; Durand *et*

al., 2016b) to generate chromatin contact maps. These maps were manually inspected and

curated using Juicebox (version 2.20.00; Durand *et al.*, 2016a) to correct misjoins and to

anchor and orient contigs into chromosome-scale scaffolds. This integration significantly

enhanced the contiguity and accuracy of the assembly.

To scaffold the primary assembly produced by hifiasm, we employed RagTag (version 2.1.0;

Alonge *et al.*, 2022), using the *A. alpina* assembly as a reference (Suppl. File 2) to further

improve scaffold ordering and correct structural inconsistencies. In addition, a linkage map

(see below) allowed us to correct a few minor assembly errors. The final assemblies have

been deposited in the European Nucleotide Archive (ENA) and is awaiting an accession

number (Project id: PRJEB89863). The ab-initio annotation was generated using the *A.*

nemorensis reference genome, which was uploaded to the PlabiPD Helixer platform

(https://www.plabipd.de/helixer_main.html). The annotation was performed in "Lineage-
Specific" mode with selection of the option "Land Plant".

Analysis of RAD-seq Data, SNP Calling, and Construction of Linkage Map

RAD-seq read quality was assessed with FastQC (version 0.11.9; Andrews, 2010). PCR
duplicates were removed using the clone_filter module in Stacks (version 2.59; Catchen *et al.*,
2013), based on a 5 bp random nucleotide sequence at the end of each adapter. Adapters
were trimmed and reads shorter than 60 bp were removed using Cutadapt (version 1.18;
Martin, 2011). We demultiplexed samples and filtered out reads with ambiguous barcodes
(allowing one mismatch), cut sites, uncalled bases, and low-quality reads (default threshold)
using the process_radtags module in Stacks (version 2.59). Reference-based mapping and
read-filtering were conducted with BWA (version 0.7.17; Li & Durbin, 2009) using default
settings against the *A. nemorensis* reference genome, along with SAMtools (version 1.10; Li
et al., 2009) and bash scripts from Rivera-Colón & Catchen (2022) (Suppl. File 3).

We performed variant-calling for 801 individuals using BCFtools mpileup and call (version
1.18; Li *et al.*, 2009) under specific criteria: a base quality score greater than 30, base quality
recalculation (-E option), and SNP calling with a significance threshold of *p*-value < 0.05.
Genotyped loci were filtered using VCFtools (version 0.1.17; Danecek *et al.*, 2011) to exclude
loci with more than 50% missing data across individuals. Only biallelic sites were kept, and
indels were removed. Individuals with over 60% missing data were excluded, leaving a
working set of 781 individuals for subsequent analysis. Additional filters were applied to site
and genotype depth (--min-meanDP 4 --max-meanDP 40 --minDP 5 --maxDP 40), and sites
with more than 70% missing data were then removed. Loci spaced less than 100 bp apart
were clustered into RAD regions as described by Dittberner *et al.* (2019). Regions with
abnormally high or low coverage were excluded based on specific thresholds: mean coverage
greater than twice or less than one-third the overall mean, maximum coverage exceeding
twice the mean maximum coverage across all regions, or regions shorter than 150 bp. Sites
absent from parental lines and ambiguous bases were also removed. Finally, we applied a
minor allele frequency filter (--maf 0.25) to the dataset. SNPs (single nucleotide
polymorphisms) were extracted using VCFtools (version 0.1.17; Danecek *et al.*, 2011) and
custom Python scripts, resulting in a VCF file containing 5,360 SNP markers across 781
individuals (Suppl. File 3).

We applied additional filters to improve the quality of map construction. We removed 39
individuals with missing data exceeding 3,500 loci. Duplicated markers were identified and
removed, resulting in 47 markers being discarded. Subsequently, markers with more than 23%

missing data were filtered out. To analyze the remaining 2,164 markers, we examined their
distribution along the genome, compared segregation distortion patterns with allelic proportion
and missingness, and filtered out markers identified as outliers based on allelic proportion. We
further removed markers with fractions of heterozygotes lower than 25% or higher than 75%.
We also used RepeatMasker (version 4.1.6; Smit *et al.*, 2015) on genome annotations to
detect and remove markers within repetitive regions. We used repeatmodeler to identifiy
repeat families, an approach that was used in Repeatmasker to mask the repetitive regions in
the genome. This final analysis produced a genetic map of 2,082 markers distributed across
eight chromosomes for 742 individuals (Suppl. File 4).

To improve the accuracy of genotype calling, particularly for heterozygous loci located in low-
coverage regions, a Python script was developed to correct incorrectly called genotypes and
impute missing data using a sliding window approach. The 2 Mb window size with a 0.5 Mb
step size demonstrated the most effective correction and imputation. No further markers or
individuals were excluded after this correction and imputation step. The finalized corrected
dataset was then used to construct the genetic map using the ASMap package in the R
environment (version 1.0-7; Taylor & Butler, 2017) packages in R (version 4.2.3), applying the
so-called Kosambi mapping function in the R/qtl package (version 1.66; Broman *et al.*, 2003)
as detailed in Supplemental File 4.

Chromosomal orientation was assessed using the constructed genetic map, identifying
inversions on chromosomes 3, 4, 6, and 7 due to assembly issues. These regions were
adjusted by inverting physical distances on the genetic map, and the genome assembly was
subsequently corrected. See above (Genome Assembly and Annotation) for details on the
genome assembly process.

Segregation Distortion and Selection Coefficient

We assessed segregation distortion within each region by applying a test of segregation
distortion (profileMark) for each marker with Bonferroni correction using the ASMap package
in R. To assess deviations from the Hardy-Weinberg equilibrium (HWE) and estimate selection
coefficients, allele frequencies for each genotype (NN, NS, SS) were calculated based on the
total number of individuals (N = 742). The observed genotype counts were used to compute
allele frequencies ($Freq_N$ and $Freq_S$) by summing homozygous and heterozygous
contributions. Expected genotype frequencies under the HWE were then computed as $Freq_N{}^2$
(NN), $2 \times Freq_N \times Freq_S$ (NS), and $Freq_S{}^2$ (SS). These frequencies were scaled to expected
counts by multiplying by the total number of individuals (Suppl. File 4).

To determine significant deviations from the HWE, a chi-squared goodness-of-fit test was

performed for each locus. The observed and expected genotype counts were compared using

the chi-squared statistic, with 2 degrees of freedom. Significant deviations were identified

based on a Bonferroni-corrected *p*-value threshold, accounting for multiple testing across all

loci. The Bonferroni correction was applied by adjusting the significance threshold to 0.05

divided by the total number of markers (2,082); this adjustment ensured that false positives

were stringently controlled for. Additionally, selection coefficients (t and s) were estimated by

calculating the relative differences in ratios of observed to expected frequencies among

genotypes. Specifically, the coefficients were defined $s_{NN} = 1 - \frac{ratio_{NN}}{ratio_{NS}}$ and $s_{SS} = 1 - \frac{ratio_{SS}}{ratio_{NS}}$,

reflecting the relative fitness difference between homozygotes and heterozygotes (Carey &

Ganders, 1980) (Suppl. File 4).

Significance of genotype deviations from the HWE was assessed using a chi-squared test for

each marker. *P*-values were adjusted for multiple testing using the false discovery rate (FDR)

correction method, controlling the FDR at 5%. Markers with FDR-adjusted *p*-values below 0.05

were considered significant and were visually highlighted in the selection coefficient plot. Red

and blue circles denote significant selection against NN (t) and SS (s), respectively. When

both t and s were significantly greater than zero, markers were interpreted as showing

potential selection favoring the heterozygote (NS).

QTL Mapping

QTL analyses were conducted using the R/qtl and QTLtools packages (version 1.66; Broman

et al., 2003; version 1.3.1; Delaneau *et al.*, 2017) in R. QTL mapping was performed on the

residuals extracted from models that accounted for positional effects. Additionally, the effect

of variation in the chloroplast and mitochondrial genomes was assessed by including cross

direction in the model alongside environmental effects. QTL mapping was conducted for 22

traits listed in Table 1. However, survival after flooding, being a binomial trait, lacked a

sufficient number of F2 individuals for QTL mapping (Suppl. File 5).

For each trait, we used the *scantwo* function to perform a two-dimensional genome scan with

a two-QTL model, applying the Haley-Knott regression algorithm (Haley & Knott, 1992).

Penalties were calculated from 1,000 permutations of the *scantwo* function to support the

stepwise fitting of multiple QTL models. The *stepwiseqtl* function was then employed, with a

maximum of five QTL, using Haley-Knott regression to search for optimized models. We

activated the *refine.locations* option in *stepwiseqtl* to improve the localization of QTLs and

deactivated the *additive.only* option to allow for potential interactions between QTLs in the

model.

We then looked into the summary output of *stepwiseqtl* to obtain information on the percentage

of variance explained by each QTL, as well as the peak LOD scores, and the additive (a) and

dominance (d) effects for each significant QTL, using allele N, identified in the best stepwise

models for each trait. For each identified QTL, we determined the 1.5 LOD confidence interval

using the *lodint* function in R/qtl. Finally, we used the *segmentsOnMap* function from QTLtools

to visualize QTL segments on the genetic map, and ggplot2 (version 3.5.1; Wickham, 2011)

to plot QTL effect sizes using the LOD scores obtained from the summary output of

stepwiseqtl. To assess the distribution of estimated additive and dominance effects, we

performed a separate Shapiro-Wilk normality test in R (Villasenor Alva & Estrada, 2009) for

each standardized effect type using standardized phenotype residuals derived from a GLM

that accounted for random effects, enabling comparisons across traits (Suppl. File 5). To

identify independent QTLs-those not associated with fertility QTLs or distortion regions-we first

excluded all QTLs located on chromosomes 4 and 7. Among the remaining QTLs, we removed

those whose genomic intervals overlapped with Fertility Score trait QTLs on the same

chromosome. Overlap was defined as any intersection between the start and end positions of

the QTL intervals.

Analysis of Whole Genome Re-sequencing Data

We used previously published resequencing data for 37 *A. nemorensis* and *A. sagittata*

individuals (Dittberner *et al.*, 2022). Additionally, we processed one *A. androsacea* individual

used as an outgroup (Dittberner *et al.*, 2022).

Reads were mapped to the new version of the *A. nemorensis* reference genome (see Genome

Assembly and Annotation) using BWA mem (version 0.7.17; Li & Durbin, 2009) and default

parameters. Read depth and other relevant read alignment quality control metrics were

computed using QualiMap v.2.2.1 (Okonechnikov *et al.*, 2016). Average read depth across all

37 samples was 23x. Variant calling was performed using GATK version 3.8 (McKenna *et al.*,

2010) and duplicates were marked using PicardTools. Filtering was based on quality

thresholds (DP<20; QD<2; MQ < 42; FS>60; SOR > 4; ReadPosRankSum < -6; MQRankSum

< -10.5). SNPs were required to be biallelic, and sites at which more than 30% of samples

were heterozygous and more than 20% missing were discarded. We polarized *A. nemorensis*

and *A. sagittata* individuals with a custom Python script using *A. androsacea* as an outgroup.

We performed phasing using the program shapeit2 (Delaneau *et al.*, 2014), assuming a

generation time of 1 year and mutation rate per generation of 7×10^{-9} and recombination rates

based on the genetic map described below. From these steps, our dataset resulted in

12,868,614 SNPs before filtering, from which we extracted 4,455,806. While curating the

assembly we discovered that segments on chromosomes 3, 4, 6, and 7 were inverted relative

to the genetic map. Rather than remapping the reads, we lifted over the VCFs themselves: for
every variant falling inside an inverted block, we mirrored its coordinate (new = start + end –
old) and reverse-complemented the REF and ALT alleles. The corrected VCF files were used
for all downstream analyses.

Genome-wide Selection Scans

Using the 37 full genome samples, we identified selective sweeps using biallelic SNPs and
the OmegaPlus software (Alachiotis *et al.*, 2012). The OmegaPlus statistics (omega) were
calculated using a grid size of 200,000 bp. We defined a minimum window size (minwin) of 50
kb and a maximum window size (maxwin) of 100 kb for computing LD values between SNPs.
Outlier omega statistics, which indicated selective sweeps had occurred, were determined
based on the genome-wide distribution of values. To minimize false positives arising from
demographic processes, the cut-off values for the omega statistics were established using
forward simulations in SLiM4 (Haller *et al.,* 2019; Haller & Messer, 2023) that were simulated
under the demographic history inferred by Dittberner *et al.* (2022). In short, the demography
consists of two pulses of interspecific gene flow between both species, one ancient pulse that
occurred directly after the species split and one recent migration event that occurred only
between the sympatric populations.

We generated 10,000 neutral datasets of 2 Mb each under the demographic history of ancient
and recent migration events that occurred between the two species, assuming a fixed
recombination rate for each simulated block (5.5e-8). The maximum omega value from each
simulated dataset was extracted, yielding a distribution of 10,000 maximum values. The 99th
percentile of this distribution that was used as the threshold to identify outlier windows
indicated selective sweeps had occurred.

To optimize sweep detection, we tested multiple combinations of grid, minwin, and maxwin
parameters (from 50 kb to 200 kb for grid size, and 50 kb to 100 kb for minwin and maxwin).
We then applied the optimized parameters to the real dataset of *A. nemorensis* and *A.*
sagittata, extracting only the sweep regions that exceeded the simulation-based threshold,
which were considered high-confidence selective sweep regions.

We further investigated the potential association between selective sweeps and phenotypic
traits by analyzing the overlap between the identified sweep regions and QTLs in *A.*
nemorensis and *A. sagittate* (Suppl. File 7). We used only high-confidence selective sweeps
in 200kb windows across the genome and overlaid these with the positions of all QTLs. We
focused on the 10 percent quantile regions around the QTL peak positions to increase
specificity.

Fine-mapping the Largest QTL

To fine-map the largest QTL, which was associated with Days to Flowering trait (first QTL on chromosome 8), we identified F2 lines that were heterozygous for the QTL of interest and homozygous for other Days to Flowering QTLs; a total of 15 lines resulted (Tables S3-4). We sowed 90 F3 seeds per line (3 seeds per pot) in trays and transferred these to the cold chamber for vernalization. After two weeks, we transplanted 1 seedling into each 7x7 cm pot filled with *Topferde* (Einheitserde, Sinntal-Altengronau, Germany), resulting in 30 seedlings per line. Pots were kept in the greenhouse for six weeks, after which plants were transplanted into larger 9x9cm pots and moved to cold frames in the garden for seven weeks. During this period, leaf material was harvested from each plant, and DNA was extracted from fresh leaves using the NucleoSpin® 8 Plant II protocol. Subsequently, the plants were transplanted into new 11x11 cm pots and placed on tables in the garden under a bird-protected cage. We recorded flowering time, inflorescence height, internode length, number of shoots, plant height, stem leaf density, stem height, and rosette diameter. Two successive trials were carried out on 483 plants, the first started on 26.09.2022, the second on 16.11.2022. Plants flowered between April and June 2023.

Geneious Prime® (version 2024.0.5) was used to design multiple species-specific primers targeting the QTL region based on the *A. nemorensis* and *A. sagittata* genomes. We designed five pairs of PCR markers, which divided the QTL region into four intervals. PCR was conducted on the DNA of 407 plants to identify recombinants and locate their recombination events within the predefined intervals. We built a quantitative model using the *glm* function from the stats package (version 3.6.2) in R, accounting for both family and trial effects as fixed factors (glm(flowering_time ~ family * (exp/tray_garden), family = quasipoisson). Then we extracted residuals and ran another model on the residuals (glm(res_ftime ~ interval1 + interval2 + interval3 + interval4, family = gaussian)). The model was run for each interval, with the interval that best explained variation in flowering time residuals containing the flowering time QTL (Suppl. File 6). The same procedure was then applied to assess the impact of each interval on traits with overlapping QTLs in the same region as the F2 mapping population (Inflorescence Height, Rosette Diameter, and Stem Height). Parental DNA and nuclease-free water were used as positive and negative controls, respectively.

Results

Phenotypic Differences of Ecologically Relevant Traits

In total, 1,193 individuals germinated in the greenhouse and were set to grow in a common garden situated at the University of Cologne. All replicates of *A. nemorensis* and *A. sagittata*

survived the common garden experiment (35 *A. nemorensis* and 35 *A. sagittata*). *A. nemorensis* flowered approximately 12 days earlier than A. sagittate, (*p*<0.001, Table 1). Mean values for 19 of the 22 scored traits differed between species (Table 1, Suppl. File 1, Table 1).

Table 1. Differences in phenotypic traits between parents. This table summarizes the phenotypic traits measured in the common garden experiment for F2 hybrids and their parental replicates (*A. nemorensis* and *A. sagittata*). Means represent the average value for each trait within parental replicates (35 individuals per species). "Distance" shows the absolute difference between the species' means. The *p*-values indicate the significance of the difference between species based on t-statistics.

Trait	Mean *A. nem*	Mean *A. sag*	Distance	*p*-value
Days to Bolting (B.T)	157.2857	169.2857	12	1.302E-19
Days to Flowering (F.T)	183.5714	193.4286	9.8571	5.556E-14
Fertility Score (W.S)	0.0298	0.0295	0.0004	>0.63
Inflorescence Height (P.H)	35.2500	34.5714	0.6786	4.537E-02
Lamina Length (Lam)	1.2729	1.4646	0.1916	1.179E-03
Lamina L/W (LLW)	1.4760	1.7938	0.3178	3.702E-12
Leaf Length (L.L)	1.5562	1.7161	0.1599	>0.75
Leaf Width (L.W)	0.8589	0.8213	0.0376	1.667E-02
Number of Stem Leaves (N.L)	28.1429	18.1429	10	4.567E-20
Petal Length (Pet)	4.0569	5.3579	1.3010	1.887E-21
Petiole Length (Pti)	0.2833	0.2516	0.0317	1.084E-06
Rosette Diameter 1 (RD1)	1.8331	2.2419	0.4087	>0.23
Rosette Diameter 2 (RD2)	3.2946	3.6893	0.3947	4.922E-03
Rosette Diameter 3 (RD3)	3.3611	3.4424	0.0813	1.502E-03
Rosette Diameter 4 (RD4)	3.8805	3.9110	0.0305	7.447E-03
Side Shoots (Ssh)	2.7857	2.1429	0.6429	2.802E-02
Stem Height (S.H)	40.4286	34.4286	6	2.668E-09
Stem Leaf Density (SLD)	0.6979	0.5271	0.1707	5.508E-16
Stem Leaf Length (SLL)	2.3894	2.7921	0.4028	1.684E-04
Stem Leaf Width (SLW)	1.4021	1.2337	0.1684	2.966E-03
Petiole L/Lamina L (P.L)	0.2464	0.1886	0.0579	5.248E-07
Ground Shoots (Gsh)	0.2143	1.2857	1.0714	1.414E-08

A. nemorensis individuals displayed a markedly higher number of stem leaves compared to *A. sagittata* (mean difference = 10 leaves, $p<0.001$), as expected for a trait that is often used to determine the taxonomy of these species (Titz, 1979). The rosette diameter of the parental genotypes differed significantly after 2, 3 and 4 weeks of growth (maximum $p<0.007$). Moreover, trait variance among F2 hybrids was generally larger than trait variance between their parental genotypes, suggesting transgressive segregation (Fig. 1). Although the seed production of parents did not differ significantly (Fertility Score or W.S, $p>0.63$), the low fertility observed for most F2 contrasted with the normal fertility of F1s (Fig. S1) and indicated outbreeding depression (Fig. 1).

After four weeks of flooding, the survival rates of parental genotypes showed no significant differences ($\chi^2(1,N=14)=0.43$, $p=0.5116$).

The cross direction impacted several traits, suggesting maternal influence and potential cytoplasmic effects (Table S1). Traits such as Stem Leaf Length (SLL, $p=3.10^{-5}$), Rosette Diameter at early stages (RD1, $p=8.10^{-5}$), and Petiole Length (Pti, $p=0.006$) exhibited strong associations with the direction of the cross. Additionally, Petiole Length-to-Lamina Length Ratio (P.L, $p=0.011$), Rosette Diameter 2 (RD2, $p=0.013$), and Rosette Diameter 3 (RD3, $p=0.03$) also displayed significant maternal effects. *A. sagittata* maternal cytoplasmic inheritance decreased rosette lamina size on the rosette, but increased it on the stem (Table S1, Fig. S2). In contrast, the majority of traits, including Days to Bolting (B.T), Days to Flowering (F.T), Fertility Score (Seed Production or W.S), and structural traits like Inflorescence Height (P.H) and Number of Stem Leaves (N.L), showed no cytotype effects ($p>0.05$), indicating limited or negligible maternal influence.

Spearman correlation analysis of 22 traits in the F2 population revealed relationships among traits (Figs. S3-4). Traits associated with leaf shape, such as Lamina Length (Lam) and Leaf Width (L.W), exhibited strong positive correlations (r > 0.8, $p < 0.001$). Similarly, Rosette Diameter (RD1–RD4) measured at different time points showed strong positive correlations (r > 0.7, $p < 0.001$).

Developmental traits, such as Days to Bolting (B.T) and Days to Flowering (F.T), showed moderate correlations with vertical growth traits, including Stem Height (S.H) and Inflorescence Height (P.H). For instance, Inflorescence Height was positively correlated with Days to Flowering (r = 0.535, $p < 0.001$), suggesting that later-flowering individuals allocate more resources to vertical growth. In contrast, Fertility Score (Seed Production or W.S), a

component of fitness, exhibited weaker and often insignificant correlations with other traits,
indicating potential independence from vegetative and structural phenotypes.

Petiole traits, including Petiole Length (Pti) and its ratio to Lamina Length (P.L), which are
often associated with shade avoidance strategies, were moderately correlated with rosette
size and leaf shape traits such as Leaf Width (L.W). Overall, these findings highlight the
interdependencies among structural traits (e.g., plant size, leaf shape), developmental traits
(e.g., flowering time, rosette size over time), and fitness components in *Arabis* F2 hybrids. The
statistical analyses of ecologically relevant traits in the common garden experiment provide
valuable insights into the extent of phenotypic integration and the nature of resource allocation
strategies in the F2 hybrid population.

Figure 1. Phenotype distribution in the F2 progeny and their parental lines. Histograms
illustrate the distribution of traits in the 1,193 plants of the F2 generation grown in common
garden. Boxplots highlight trait variations between the two parental lines and indications of
transgressive segregation in their offspring. The *p*-values and stars depict the significance of
trait differences between parents * p<0.05, ** p<0.01, *** p<0.001. Abbreviations are as
follows: Days to Bolting (B.T), Days to Flowering (F.T), Fertility Score - Seed Production (W.S),

Inflorescence Height (P.H), Lamina Length (Lam), Lamina Length-to-Width Ratio (LLW), Leaf
Length (L.L), Leaf Width (L.W), Number of Stem Leaves (N.L), Petal Length (Pet), Petiole
Length (Pti), Rosette Diameter at four time points (RD1-4), Side Shoots (Ssh), Stem Height
(S.H), Stem Leaf Density (SLD), Stem Leaf Length (SLL), Stem Leaf Width (SLW), Petiole
Length-to-Lamina Length Ratio (P.L), and Ground Shoots (Gsh).

Genetic Map, Pre- and Post-zygotic Selection Distortion

To investigate the genetic architecture underlying trait variation, we used a reduced
sequencing approach to determine the genotypes of 742 F2 individuals at 2,082 reliable SNP
markers with the help of high-quality genome assembly (Fig. S5, Suppl. File 2, Table 1), and
constructed a genetic map. The map consisted of eight linkage groups (LGs), which were
numbered according to the chromosome numbers of *Arabis alpina*. The length of the linkage
map was 240 cM, with 160 to 369 markers per chromosome (Table S2). Contrasting the
genetic and physical distances of SNPs along chromosomes showed that recombination was
higher on chromosome arms compared to in centromeric regions (Figs. S6-7).

The analysis of allele and genotype frequencies along the genome uncovered strong
segregation distortions in the F2 population. A strong segregation distortion was found on
chromosomes 4 and 7 with the percentage of N alleles dropping from 50% (expected) to 32%
(Fig. 2A). Because the sequencing data was mapped on the high-quality *A. nemorensis*
genome, we can conclude that this depletion is clearly not due to mapping biases. To
disentangle prezygotic from postzygotic mechanisms of allele distortion, we estimated the
difference between expected and observed genotype distributions across the genome using
a chi-square test (Fig. 2B), and inferred selection coefficients against each type of
homozygote. This analysis showed that the segregation distortions on chromosome 7 and on
the tip of chromosome 4 were not due to postzygotic selection on the genotype. Yet, it
identified 6 regions in the genome with a significant excess of heterozygotes, three of which
were particularly strong on chromosomes 1 and 8. It further highlighted one region in
chromosome 3 with massively depleted homozygotes.

The extent of segregation distortion found throughout the genome appeared to be mostly due
to interactions between local alleles. For example, no genetic association was observed
between the distortion on chromosomes 4 and 7 (Fig. S8). Nevertheless, we find biased
transmissions of parental alleles on chromosomes 2 and 5, as well as chromosomes 3 and 8
(Fig. S8).

Figure 2. Segregation distortion, genotypic distortion and strength of selection along the genome. This figure illustrates segregation distortion, genotypic distortion, and the strength of selection across the genome in the F2 population. The x-axis represents marker positions along the genome across the eight chromosomes (2082 markers). The y-axis in (A) and (B) shows $-\log_{10}(p\text{-value})$; the y-axis in (C) represents the selection coefficient. (A) Gametic distortion: shown by the deviation of SNPs from expected Mendelian segregation ratios, assessed using a segregation distortion test (profileMark) with Bonferroni correction for multiple testing (threshold $p < 2.4 \times 10^{-5}$). Significant deviations are highlighted in red. 1,257 markers are significantly distorted. (B) Genotypic distortion: calculated as the deviation of observed genotypes from expected values according to the HWE, assessed using a chi-square test. Bonferroni-adjusted significant differences are highlighted in red. Only 47 markers surpassed the Bonferroni significance threshold. (C) Selection coefficient (s): calculated based on deviations from expected allele frequencies, with selection on allele S represented in green and on allele N in purple. Red and blue circles indicate markers where genotype frequencies significantly deviate from the HWE (FDR-adjusted $p < 0.05$, chi-squared test), suggesting selection at those loci. Red circles represent significant selection against the NN genotype (t); blue circles represent selection against the SS genotype (s). Loci where both t and s are significant and positive, indicate potential selection favoring the heterozygote (NS).

Genetic Architecture of Ecologically Relevant Traits

We detected significant QTLs for 20 of the 22 traits we scored (Table 2). The number of QTLs per trait ranged from 1 (Lamina Length) to 5 (Days to Flowering). The most significant QTL (LOD score > 40) was identified on chromosome 8, accounting for more than 20% of the variation in flowering time. In contrast, the lowest LOD score (3.027) was observed in one of the Leaf Width QTLs. In total, 58 QTLs were identified along the genome (Table 2). We observed overlapping QTLs for multiple traits, including 9 on chromosome 1; 3 on chromosome 2; 11 on chromosome 3; 4 on chromosome 4; 3 on chromosome 5; 2 on chromosome 6; 6 on chromosome 7; and 20 on chromosome 8 (Fig. 3). Notably, one QTL associated with the Fertility Score on chromosome 6 did not overlap with any other QTLs (Fig. 3, Table 2)

Figure 3. Genetic map with QTLs for ecologically relevant traits. The plot above illustrates significant QTLs that were detected and located on the genetic map. Horizontal bars represent mapped SNP markers. Gaps between bars stand for the genetic distance between SNP markers in cM (centimorgan). Traits are listed in alphabetical order. The chromosome containing the strongest QTL of each trait appears in the square.

The strongest QTLs for traits such as Days to Bolting, Days to Flowering, Inflorescence Height,
Leaf Length and Width, Number of Stem Leaves, and Rosette Diameter after 1, 2, and 4
weeks, as well as Stem Height and Stem Leaf Width, were all located on chromosome 8.
Additionally, strongest QTLs for Petal Length, Rosette Diameter 3, Side Shoots, Stem Leaf
Density, and Stem Leaf Length were identified on chromosomes 7, 1, 5, 3, and 6, respectively.
For all traits, the additive effect of the *A. nemorensis* allele was either positive or negative,
confirming that transgressive genetic variation can arise in most traits in this selfing species
by fixing a new combination of alleles (Table 2).

The distribution of fertility scores, which was measured as the weight of seeds contained in
10 siliques, indicated that F2 individuals tended to be less fertile than the parental lineages.
The genetic architecture of fertility score variation was dominated by a large-effect QTL (LOD
= 8.774) on chromosome 3 (Fig. S9), which explained more than 30% of the phenotypic
variation and involved inter-allelic incompatibility at position 8,362,080 bp on chromosome 3
(QTL interval: 6,744,303 to 9,601,991 bp, Fig. S9). Individuals with the NS heterozygous
genotype at this marker displayed markedly lower fertility. In addition, three smaller QTLs
explaining 2.683%, 4.641%, and 3.507% of the variation were found on chromosomes 6, 7,
and 8, respectively. It can be concluded that the genetic basis of outbreeding depression is
relatively simple in this population.

Figure 4. Distribution of standardized QTL estimated additive and dominance effects.
The figure above displays the distribution of standardized estimated additive and dominance
effects for QTLs associated with ecologically relevant traits identified in the F2 population.

Each curve represents the density of standardized effect sizes for a distinct type of genetic
effect: additive effects (in light blue) and dominance effects (in purple). The yellow diamond
marks the dominance effect of the strongest QTL for Fertility Score (W.S) on chromosome 3.
Effect values were estimated using interval mapping on standardized phenotype residuals
derived from a GLM that accounted for random effects, enabling comparisons across traits.
Density distributions illustrate the range and frequency of QTL contributions. Additive effects
showed a Gaussian distribution (Shapiro-Wilk: $W = 0.974$, $p = 0.257$), whereas dominance
effects approximated a logistic distribution after excluding the outlier dominance effect of a
QTL on chromosome 3 (Shapiro-Wilk: $W = 0.683$, $p < 0.001$).

Of the 22 traits scored, Ground Shoots (Gsh), Petiole Length-to-Lamina Length ratio (P.L),
and Survival to Flooding revealed no significant QTL. For two of these three traits (Ground
Shoots, Petiole Length-to-Lamina Length ratio), significant differences between species were
observed. The absence of QTLs for these traits suggests they may have a polygenic genetic
basis, and the effect of individual variants may be too small to be detected.

We conclude that the two parental lineages differ genetically in many traits, with size effects
of various sizes. The number of detected QTLs allows us to describe the size effects of genetic
variation that can be generated by hybridization in this system. The standardized additive
effects were normally distributed (Shapiro-Wilk: $W = 0.9744$, $p = 0.2573$, Fig. 4), ranging from
a minimum of -0.15 to a maximum of 0.20 with a median of 0.008 (mean = 0.005, s.d.= 0.08,
skewness = 0.29, kurtosis = 2.6). QTLs with the largest effect controlled flowering time, stem
height, inflorescence height, leaf number and lamina shape contributing to the heavy tail of
the distribution of additive effect (Figs. 4, S10). Standardized dominance effects instead
approximated a logistic distribution (Shapiro-Wilk: $W = 0.683$, $p < 0.001$), ranging from -0.36
to 0.11 with a median of 0.012 (mean = 0.008, s.d. = 0.062, skewness =-3.70, kurtosis = 23.5).
The dominance effect of the fertility score QTL on chromosome 3 was an outlier to this
distribution (Fig. 4).

Table 2. QTLs of ecologically relevant traits in the *Arabis* mapping population. This table
summarizes the number, chromosomes, peak positions, LOD scores, percentages of
phenotypic variance explained, and estimated additive and dominance effects of allele S for
each QTL. The number of observations per trait in the F2 population is indicated. Significance
differences between parental lines are denoted as follows: *** ($p < 0.001$), ** ($0.001 < p < 0.01$), * ($0.01 < p < 0.05$), . ($0.05 < p < 0.1$) (details in Suppl. File 5)
0.01), * ($0.01 < p < 0.05$), . ($0.05 < p < 0.1$) (details in Suppl. File 5)

Trait	No. of Observations	Significance ofParental Differences	Chr	Position (bp)	Position (cM)	LOD	%Variance Explained	est *a*	est *d*
Days to Bolting (B.T)	631	***	3	20040893	163	5.146	3.130	0.005914	-0.012765
			7	1835566	21.5	5.359	3.263	0.010990	0.002648
			8	2061023	12.2	7.786	4.783	-0.011989	0.000447
			8	19500950	134.2	12.396	7.746	0.016244	-0.003620
Days to Flowering (F.T)	578	***	3	26882840	190	9.915	4.785	0.012242	-0.001815
			5	23418119	111	4.868	2.289	-0.007092	0.003427
			7	1835566	21.5	11.241	5.423	0.011243	-0.001854
			8	2061023	13	42.112	23.075	-0.021252	0.008320
			8	18832553	133	9.374	4.488	0.009159	-0.005437
Fertility Score (W.S)	552		3	8362080	63	8.774	6.165	-0.056270	-0.366350
			6	1052484	10	3.897	2.683	0.172390	-0.055580
			7	3192272	31	6.663	4.641	0.205690	-0.053080
			8	3118527	17	5.070	3.507	0.178080	-0.028820
Inflorescence Height (P.H)	578	.	1	9004320	49.9	4.166	2.385	-0.012682	0.062053
			3	2664384	27	11.394	6.714	0.078301	0.016361
			8	1180103	6	20.789	12.728	-0.094477	0.076771
			8	10128705	101	6.383	3.686	-0.062798	-0.038649
Lamina Length (Lam)	436	*	8	5481058	36.0	4.334	4.475	-0.075590	0.042250
	436	***	3	6744303	61	15.635	14.890	0.089464	-0.031374

Trait	No. of Observations	Significance ofParental Differences	Chr	Position (bp)	Position (cM)	LOD	%Variance Explained	est *a*	est *d*
Lamina L/W (LLW)			4	38897287	163	4.045	3.620	0.041307	0.028487
Leaf Length (L.L)	436		1	13071968	72.4	4.458	3.766	0.071867	-0.012244
			3	14255522	136.2	5.214	4.423	-0.007829	0.118366
			3	29031007	201.4	5.617	4.775	0.066584	-0.094772
			7	14342773	68.8	4.284	3.615	-0.020747	0.086727
			8	5624776	35	9.433	8.183	-0.104748	0.036218
Leaf Width (L.W)	436		1	21722429	116	4.313	3.893	0.069770	0.045460
			2	19435694	87	3.027	2.713	-0.043950	0.056000
			2	25576894	129	5.987	5.452	0.091250	0.012400
			8	4356973	32	6.705	6.129	-0.081970	0.036770
Number of Stem Leaves (N.L)	625	***	3	14075550	130.9	4.581	2.557	0.025147	0.059693
			4	7558305	18	4.001	2.228	0.056356	0.010451
			5	16624270	70.7	5.384	3.014	-0.055649	0.004690
			8	1180103	7	28.109	17.139	-0.119377	0.052082
Petal Length (Pet)	106	***	1	9074627	51	4.863	15.820	0.064910	0.039080
			7	1835566	21.5	5.689	18.850	0.079160	0.013480
Petiole Length (Pti)	361	**	8	10723493	107.5	3.814	4.749	-0.153200	-0.035130
Rosette Diameter 1 (RD1)	695		1	17879753	95	4.498	2.747	0.068440	-0.020590
			3	623617	2.3	3.632	2.212	0.054350	0.023100
			8	4362085	28	6.942	4.275	-0.074250	0.032990
Rosette Diameter 2 (RD2)	700	.	1	21954996	115	6.158	3.703	0.079730	0.013670
			7	5769578	42.7	3.839	2.291	-0.056060	0.024000
			8	4356973	31	7.822	4.729	-0.085450	0.033910
	696	.	1	9004320	49.9	9.389	5.683	0.102710	-0.005208

Trait	No. of Observations	Significance of Parental Differences	Chr	Position (bp)	Position (cM)	LOD	%Variance Explained	est *a*	est *d*
Rosette Diameter 3 (RD3)			2	23541930	118	4.024	2.392	0.043216	0.071859
			8	2718077	18	7.209	4.332	-0.087105	0.022622
Rosette Diameter 4 (RD4)	701	*	1	21722429	115.3	4.330	2.678	0.061815	0.014845
			8	4356973	32	7.814	4.888	-0.078677	0.035138
Side Shoots (Ssh)	595	*	4	36501835	137.7	3.859	2.842	-0.119930	-0.038550
			5	5129643	35.5	4.875	3.605	0.126580	0.031450
Stem Height (S.H)	625	***	1	29847516	162	3.594	1.778	-0.029247	0.036282
			3	4035375	36	13.598	6.984	0.074366	0.005981
			4	38897287	160	4.386	2.177	0.042699	-0.004189
			8	1180103	6	26.851	14.500	-0.091994	0.070533
			8	10506892	102	7.858	3.951	-0.059677	-0.004408
Stem Leaf Density (SLD)	624	***	3	3491752	29	5.760	4.162	-0.052426	-0.018081
Stem Leaf Length (SLL)	259	**	6	21070306	134.8	5.918	9.356	0.046349	0.011155
			8	3118527	15	3.967	6.163	-0.035380	0.011240
Stem Leaf Width (SLW)	259	***	8	18832576	132.5	13.228	20.958	-0.084603	0.017725

Genes Responsible for Flowering Time

The largest QTL detected in this study impacts the timing of flowering and is located on chromosome 8, where two well-known genes that control flowering time are found: *FLC*, a regulator involved in the vernalization pathway, and *CONSTANS*, a regulator involved in the photoperiod pathway (Andrés & Coupland, 2012). *FLC* has been shown to be a key variant shaping flowering time in many Brassicaceae species, including the congeneric species *Arabis alpina* (Albani *et al.*, 2012). The gene *CONSTANS* is also located within the boundaries of this QTL. In order to test whether one of these candidate genes was responsible for the variation, we selected 15 F2 individuals that were heterozygous in the QTL region of chromosome 8 and homozygous on the other QTLs, and we grew 30 of their seeds. A total of

410 plants were assessed for flowering time in 15 F3 families and two trials (Fig. S11, Tables
S3-4).

With 410 plants and a QTL region that was ~17 cM, we expected 80 recombinants.
Interestingly, we identified 138 recombinants, suggesting that the recombination rate had been
slightly underestimated in the F2 population. We compared four models to identify the
chromosomal fragments that best explained variation in flowering time, after accounting for all
other factors of the experimental design. Using Akaike's criterion and comparing p-values, we
identified fragment 2 as the most likely to contain the variants causing a difference in flowering
time (Fig. 5, p=0.0118). This segment ranges from position 1,831,324 bp to position 2,125,083
bp at the beginning of chromosome 8. Although this ~300 kb region does not contain *FLC* or
CO, it contains 64 other genes, 31 of which have a known ortholog in *Arabidopsis thaliana*.
Only 1 of these, *TFL1*, is known to regulate flowering time in *A. thaliana* and *A. alpina*, with a
loss of function inducing early flowering in both species (Cerise *et al.*, 2023). *A. sagittata TFL1*
differs from the *A. nemorensis* copy by 2 amino acid exchanges: Asparagine 3 is changed to
Isoleucine and Serine 50 is changed Tyrosine. The *A. nemorensis* version at both sites is
conserved in most Brassicaceae, like *A. thaliana*, where *TFL1* function has been
experimentally verified (Andrés & Coupland, 2012). Nevertheless, both positions are not
strictly conserved.

In contrast to most *TFL1* genes from the Brassicales, several species within (*Capsella rubella*,
Sinapis alba) and outside of this group are missing the start codon and protein translation
starts at the conserved Methionine codon at position 4. At position 50, Serine is the most
common amino acid not only in the Brassicales group, but also in other plant groups, but other
amino acids (Alanine, Threonine, Phenylalanine) are found in other Brassicaceae *TFL*1
sequences. Because *A. sagittata* carries derived amino acids at two moderately conserved
positions, gene activity may be decreased compared to the more ancestral *A. nemorensis*
allele. Fine-mapping, however, allowed us to exclude the role of flowering loci such as *FLC* or
CO that locate on the same chromosome arm of chromosome 8.

Although there were QTLs for Inflorescence Height, Rosette Diameter, and Stem Height, on
chromosome 8 in F2s, the chromosomal region that was fine mapped did not explain the
variation of these traits in F3s. Inflorescence Height, Rosette Diameter and Stem Height are
thus controlled by QTL(s) independent of the QTL for Flowering Time in the *TFL1*-containing
fragment, with the S allele advancing flowering compared to the N allele.

Figure 5. Association between 4 DNA fragments and flowering time within 15 F3 families segregated for the chr8 flowering time QTL. The figure illustrates the role of genomic intervals within the QTL region in explaining variations in Flowering Time, Inflorescence Height, Rosette Diameter, and Stem Height. The x-axis represents region of the strongest Flowering Time QTL, divided into intervals defined by species-specific primers with adjusted lengths for fine mapping. The dashed line indicates the significance threshold. Different bar colors represent different traits. The *TFL1* gene was detected in interval 2, which shows the strongest effect on flowering time.

Overlap of Parental QTLs with the History of Selection in the Parental Lineages

In order to test whether genetic differences between lineages were influenced by selection for ecological specialization in the two parental lineages, prior to hybridization, we used previously published population genomics data to determine regions with signature of selective sweeps (Dittberner *et al.*, 2022). We detected 35 and 82 genome regions carrying signatures of selective sweeps in *A. nemorensis* and *A.sagittata*, respectively, which prompted us to ask whether variable traits were disproportionately targeted by natural selection in their lineage or origin (Fig. S12). Many of the 58 QTL regions spanned large parts of chromosomes, sometimes overlapping with multiple independent selective sweeps (Fig. S13). We tested

whether the distance between QTL peak and selective sweep was shorter than expected by
a permutation test. QTL peaks did not tend to be located close to selective sweep signatures.
We also did not detect significant overlaps between regions with distorted segregation and
regions carrying selective sweeps (not shown). The genetic variation segregating in the F2
population thus does not appear to reflect the history of selection in the hybridizing lineages,
yet some individual QTLs may reflect recent selection. For example, sweeps overlapped with
5 of the 7 genomic regions with QTLs for rosette size (Fig. S12). In contrast, no signature of a
putative selective sweep was found in the *TFL1* region.

Discussion

Hybridization between species is pervasive in both past and contemporary ecosystems
(Lewontin & Birch, 1966; Stull *et al.*, 2023). By bringing together alleles that contribute to
different ecological specialization, it can facilitate the emergence of a new allelic combination
that may surpass the performance of the parental genotypes and be fixed via selfing.
Hybridization may therefore become pivotal in the rescue of endangered selfing species with
very low levels of genetic diversity (Frankham, 2015). Here, we take advantage of a
hybridization hotspot we identified previously to disentangle the genomic and ecological
properties of genetic variation released after hybridization, some of these properties may
propagate in the population via selfing. An F2 generation was obtained by crossing individuals
representative of the two species in the sympatric population where they naturally hybridize
(Dittberner *et al.*, 2019; Dittberner *et al.*, 2022).

The genetic analysis of allelic transmission in the F2 population first revealed the genomic
barriers that skew allele segregation in hybrid offspring. Indeed, several regions of the genome
displayed highly distorted transmission. For the two most strongly distorted regions on
chromosomes 4 and 7, the genotypic composition of the F2 fits Mendelian expectations,
suggesting that the distortion happens before fertilization, as a result of biased gamete
formation. Both of these distortions drive the *A. sagittata* allele to high frequencies in the F2
offspring population. The predominance of one allele among mature gametes produced in a
heterozygous individual represents a very strong evolutionary force called the meiotic drive
(Sandler & Novitski, 1957; Pinkas *et al.*, 1985; Clarke *et al.*, 2004; Domínguez *et al.*, 2005).
The meiotic drive can massively accelerate the fixation of alleles linked to the driver locus.
Driver alleles may outcompete non-driver alleles during pollen tube growth in heterozygotic
selfing plants, thereby biasing the paternal transmission of alleles in favor of the offspring (e.g.,
Snow & Spira, 1996; Aronen *et al.*, 2002; Lankinen *et al.*, 2009). In female gametes, any
molecular change that favors allele transmission into the polar body would lead to preferential
maternal transmission (Fishman & Kelly, 2015; Finseth *et al.*, 2015). Here, since genotype

frequencies fulfil the HWE, the driver alleles should outcompete the non-driver alleles in both

male and female gametogenesis (Malik, 2005). Many mechanisms can lead to gametic drive

in plants, because gametes undergo haploid cell division, a step that does not exist in animals

(Finseth, 2023). As a consequence, about 60% of plant genes are expressed in the haploid

phase (Chettoor *et al.*, 2014; Rutley & Twell, 2015; Klepikova *et al.*, 2016), and variants have

been identified in genes controlling cell division in gametes of *A. thaliana* (Parker *et al.*, 2025).

In addition to segregation distortion at the gametic level, we also detected regions in the

genome that experienced biased transmission as a result of the distorted frequency of

heterozygotes. One locus on chromosome 3, for example, was markedly depleted in

heterozygotes, despite equal transmission of the two alleles. This locus was close, but distinct

from a second locus that decreased the seed production of heterozygotes by 30%. This

pattern, which is also known as allelic underdominance, characterizes the genome of a pair of

alleles in *A. thaliana* (Smith *et al.*, 2011). The causal variant mapped to structural changes in

a tandem array of duplicated genes and altered kinase activity (Smith *et al.*, 2011). In selfing

species and their hybrids, underdominance will not pose a significant threat to the equal

transmission of alleles provided that F1 hybrids are viable – as is the case in this study

system– because half of the descendants of heterozygote individuals will return to

homozygosity.

Allelic underdominance is a form of Dobzhansky-Muller (DM) incompatibility (Dobzhansky,

1936; Muller, 1942). DM incompatibilities arise after alleles are fixed in the diverging parental

lineages without being naturally selected to function together. Interactions between alleles

segregating at different loci can also cause DM incompatibilities, but, interestingly, no

segregation distortions were detected in any major interlocus in this system, nor did any

interactive QTLs determine fertility. DM incompatibilities have been studied extensively in

many taxa (Bomblies & Weigel, 2007; Masly & Presgraves, 2007; White *et al.*, 2011; Schumer

et al., 2014; Zuellig & Sweigart, 2018; Coughlan & Matute, 2020). Although such

incompatibilities are assumed to evolve completely neutrally, they can also result from local

adaptation (Alcázar *et al.*, 2014).

However, whereas only 2 loci showed underdominance, 6 loci showed selection for

heterozygote genotypes. This discrepancy indicates that in this system the compensation of

deleterious variants in heterozygote regions is more predominant than the emergence of DM

incompatibilities (Clo *et al.*, 2021). Indeed, 6 regions in the genome show an excess of

heterozygotes. This imbalance indicates that the low effective population sizes of the two

species, as well as their relatively recent origin, will have increased the fixation of deleterious

alleles faster than they will have fixed new incompatible mutations (Simons & Sella, 2016;

Dittberner *et al.*, 2022). As recombination and selfing will proceed in further generations after
hybridization, high-fitness individuals are expected to arise that will have purged these
variants. The interplay between purging and recombination has been shown to cause
heterogeneous rates of introgression along the genome of hybridizing taxa (Schumer *et al.*,
2018).

The pattern of past introgression in the *Arabis* system is in fact heterogeneous along the
genome (Dittberner *et al.*, 2022). Yet, our work shows that the selective removal of deleterious
variants is probably not the only force affecting this system today. The massive transmission
advantage of the *A. sagittata* alleles on chromosomes 4 and 7 implies that any allele that is
located close to the driver of the distortion will readily introgress into the local *A. nemorensis*
population, whether it is deleterious or ecologically relevant. Linked QTLs of small effect in the
vicinity of the distortion, however, will be hard to detect. Other approaches, such as, for
example, transcriptome analyses, are needed to determine the potential ecological relevance
of genetic variation hitchhiking with gametic driver alleles.

The genomics of allele transmission is clearly shaping the consequences of hybridization in
this system. Yet, segregation distortions, incompatibilities, and overdominance were
detectable on a limited number of loci. Due to their simple genetic basis, the loci are unlikely
to block genetic exchanges (Li *et al.*, 2022). The genetic architecture of the comprehensive
panel of traits quantifying growth rate, growth form, the transition to flowering, plant height and
the production of seeds allowed us to examine the extent to which gene flow at ecologically
relevant QTLs was hindered by genomic barriers in this system. Results demonstrated that
the parental lineages differ genetically in most traits, with about 48% of the QTLs contributing
to this variation, that is, most traits are independent of the loci affecting segregation or fertility
in the F2 generations (Table S5).

Hybrid F2 offspring often exhibited phenotypic values beyond the range of the parental
lineages, particularly for plant height, rosette diameter, leaf length and flowering time. Indeed,
although the two parental genotypes did not differ in the timing of flowering, it was for this trait
that the largest QTL was detected on chromosome 8. Our study showed that changes in
flowering time in *A. nemorensis* and *A. sagittata* hybrids are not linked to the well-known genes
controlling flowering time, *FLC* or *CO* (reviewed in Alonso-Blanco *et al.*, 2009; Andrés &
Coupland, 2012). Instead, the fine-mapping of the largest QTL defined a genetic region that
contains only one candidate gene for flowering time. This gene, first described as *TFL1* in *A.
thaliana,* accelerates flowering by decreasing the number of vegetative side buds in this
species (Alvarez *et al.*, 1992; Moraes *et al.*, 2019). In the perennial species *A. alpina*, it further
interacts with the age pathway to prevent juvenile plants from responding to vernalization,

thereby controlling the transition out of the juvenile stage, an additional developmental
transition of central importance for perennials (Wang & Albani *et al.*, 2011). Because this
flowering time variant is independent from all segregation distortions and fertility alleles, it
appears as a potential adaptive trait likely to be reshaped as a result of gene flow. Studies of
flowering time provided some of the most remarkable examples of contemporary adaptation
to climate change in plants (Hancock *et al.*, 2024 submitted), and adaptive introgression of
flowering time alleles have been documented in several cases (Le Corre *et al.*, 2002; Todesco
et al., 2020; Wang *et al.*, 2021). Environmental responses to global climate change include
advancing flowering time: many plants induce earlier flowering as a strategy to escape warmer
temperatures (e.g., Anderson *et al.*, 2012; Cai *et al.*, 2019; Siegmund *et al.*, 2016; Tun *et al.*,
2021).

Which additional, genetically different traits may be adaptive in natural communities remains
to be determined. Larger leaves, greater lateral spread and lower specific leaf area, all of
which are variable in the descendants of hybrid individuals in this system, have been
associated with competitive advantage in the ephemeral wetland of vernal pool plant
communities (Kraft *et al.*, 2015). Interestingly, the *A. nemorensis* maternal background
increased rosette diameter, especially at early stages, while the *A. sagittata* maternal
background increased stem leaf length. Both of these traits can improve light capture, but their
relative fitness relevance should depend on the trade-off between survival and competition
(Lundgren & Des Marais, 2020). Backcrossing of *A. nemorensis* pollen on *A. sagittata* mother
plants appeared to occur more frequently among hybrids (Dittberner *et al.*, 2019; Dittberner *et
al.*, 2022), which would thus promote an increase in stem leaf length. Therefore, genetic and
maternal variants may contribute to the emergence of a particularly competitive genotype in
the dense plant communities of floodplain meadows. In this context, it is intriguing that
signatures of putative selective sweeps in the parental species were found for many of the
QTLs determining variation among rosette size. Recent selection in the parental lineages
however, has not been systematically associated with variable QTLs, so that much of the
variation manifested in the F2 generation forms a new base of variants that has not yet been
shaped by natural selection.

As the climate changes and land-use intensifies, the native floodplain habitat is increasingly
exposed to drought events as well as flooding, both of which will affect plant survival in natural
environments and neither of which could be quantified in this common garden experiment
(Colloff *et al.*, 2016). In situ analyses of the performance of hybrid offspring are therefore
needed to shed light on the genetic basis of variation in survival rates. We conclude that there
is sufficient genetic diversity in our system for hybridization to generate novelty. The prospects

of hybridization between these species, however, will depend on how selective pressures act
on genetically variable traits within the architecture of allelic transmission.

Acknowledgments

This work was supported by the Deutsche Forschungsgemeinschaft (DFG [ME2742/13-1,
TRR341 and EXC 2048/1–390686111). We thank the Cologne Center for Genomics (CCG),
the Regionales Rechenzentrum der Universität zu Köln (RRZK), the German Network for
Bioinformatics Infrastructure (de.NBI) and ELIXIR-DE (Forschungszentrum Jülich and W-
de.NBI-001, W-de.NBI-004, W-de.NBI-008, W-de.NBI-010, W-de.NBI-013, W-de.NBI-014, W-
de.NBI-016, W-de.NBI-022). We thank K. Schneeberger, B. Hüttel, A. Hancock, M. Stetter, I.
Calic and M. Hajheidari for valuable feedback.

Competing interests

836 None declared.

Author contributions

N.R. designed and performed the RADseq and QTL mapping analyses, conducted flowering-
time experiments and analyses, and wrote the manuscript. L.M. contributed to the whole-
genome sequencing (WGS) analyses and wrote the WGS section of the manuscript. L.H.
assisted with RADseq data analysis. K.K. assisted with genotype correction. A.S.K. generated
the genome annotations. T.A., C.I.D., S.A., R.Y.W, and N.R. generated the genome
assemblies. G.S. and B. S. contributed to data interpretation. A.T. and J.d.M. conceived and
supervised the project, contributed to data interpretation, and reviewed and edited the
manuscript.

Data Availability

FASTQ files of the F2 population, genome assemblies, and annotations are available at the
European Nucleotide Archive (ENA) under project number [PRJEB89863]. All supplementary
figures can be found at Supplementary Figures. All scripts, VCF file, and phenotypic datasets
used in this study are available on GitHub via the following links:

- Suppl. File 1. Phenotypic analyses:

https://github.com/nedarahnama/Contemporary_hybridization/tree/master/01_phenotypic
_analyses

- Suppl. File 2. Genome assembly:

https://github.com/nedarahnama/Contemporary_hybridization/tree/master/02_genome_a
ssembly

- Suppl. File 3. RADseq analysis:

858 https://github.com/nedarahnama/Contemporary_hybridization/tree/master/03_rad_seq

- Suppl. File 4. Genetic map construction:

860 https://github.com/nedarahnama/Contemporary_hybridization/tree/master/04_genetic_m
861 ap

- Suppl. File 5. QTL mapping analysis:

863 https://github.com/nedarahnama/Contemporary_hybridization/tree/master/05_qtl_mappin
864 g

- Suppl. File 6. Flowering time fine-mapping analysis:

866 https://github.com/nedarahnama/Contemporary_hybridization/tree/master/06_fine_mappi
867 ng

- Suppl. File 7. Sweep detection:

869 https://github.com/Luker121/OverlapSweepQTL/tree/main

References

References

Abbott RJ. **2017**. Plant speciation across environmental gradients and the occurrence and nature of
hybrid zones. *Journal of Systematics and Evolution* **55**.

Abbott R, Albach D, Ansell S, Arntzen JW, Baird SJE, Bierne N, Boughman J, Brelsford A, Buerkle CA,
Buggs R, *et al.* 2013. Hybridization and speciation. *Journal of Evolutionary Biology* **26**.

Alachiotis N, Stamatakis A, Pavlidis P. **2012**. OmegaPlus: A scalable tool for rapid detection of
selective sweeps in whole-genome datasets. *Bioinformatics* **28**.

Albani MC, Castaings L, Wötzel S, Mateos JL, Wunder J, Wang R, Reymond M, Coupland G. **2012**.
PEP1 of *Arabis alpina* Is Encoded by Two Overlapping Genes That Contribute to Natural Genetic
Variation in Perennial Flowering. *PLoS Genetics* **8**.

Alcázar R, von Reth M, Bautor J, Chae E, Weigel D, Koornneef M, Parker JE. **2014**. Analysis of a Plant
Complex Resistance Gene Locus Underlying Immune-Related Hybrid Incompatibility and Its
Occurrence in Nature. *PLoS Genetics* **10**.

Alonge M, Lebeigle L, Kirsche M, Jenike K, Ou S, Aganezov S, Wang X, Lippman ZB, Schatz MC, Soyk
S. **2022**. Automated assembly scaffolding using RagTag elevates a new tomato system for high-
throughput genome editing. *Genome Biology* **23**.

Alonso-Blanco C, Aarts MGM, Bentsink L, Keurentjes JJB, Reymond M, Vreugdenhil D, Koornneef M.
2009. What has natural variation taught us about plant development, physiology, and adaptation?
Plant Cell **21**.

Alvarez J, Guli CL, Yu X -H, Smyth DR. **1992**. terminal flower: a gene affecting inflorescence
development in *Arabidopsis thaliana*. *The Plant Journal* **2**.

Anderson JT, Panetta AM, Mitchell-Olds T. **2012**. Evolutionary and Ecological Responses to
Anthropogenic Climate Change. *Plant Physiology* **160**: 1728–1740.

Anderson E, Stebbins GL. **1954**. HYBRIDIZATION AS AN EVOLUTIONARY STIMULUS. *Evolution* **8**.

Andrés F, Coupland G. **2012**. The genetic basis of flowering responses to seasonal cues. *Nature*
Reviews Genetics **13**.

Andrews S. **2010**. FastQC: a quality control tool for high throughput sequence data. 2010.
Https://Www.Bioinformatics.Babraham.Ac.Uk/Projects/Fastqc/.

Aronen T, Nikkanen T, Harju A, Tiimonen H, Häggman H. **2002**. Pollen competition and seed-siring
success in *Picea abies*. *Theoretical and Applied Genetics* **104**.

Becker M, Gruenheit N, Steel M, Voelckel C, Deusch O, Heenan PB, McLenachan PA, Kardailsky O,
Leigh JW, Lockhart PJ. **2013**. Hybridization may facilitate in situ survival of endemic species through
periods of climate change. *Nature Climate Change* **3**.

Blanckaert A, Sriram V, Bank C. **2023**. In search of the Goldilocks zone for hybrid speciation II: hard
times for hybrid speciation? *Evolution* **77**.

Bomblies K, Weigel D. **2007**. Hybrid necrosis: Autoimmunity as a potential gene-flow barrier in plant
species. *Nature Reviews Genetics* **8**.

Bontrager M, Angert AL. **2019**. Gene flow improves fitness at a range edge under climate change.
Evolution Letters **3**.

Brauer CJ, Sandoval-Castillo J, Gates K, Hammer MP, Unmack PJ, Bernatchez L, Beheregaray LB.
2023. Natural hybridization reduces vulnerability to climate change. *Nature Climate Change* **13**.

Broman KW, Wu H, Sen Ś, Churchill GA. **2003**. R/qtl: QTL mapping in experimental crosses.
Bioinformatics **19**.

Buerkle CA, Morris RJ, Asmussen MA, Rieseberg LH. **2000**. The likelihood of homoploid hybrid
speciation. *Heredity* **84**.

Burmeier S, Eckstein RL, Donath TW, Otte A. **2011**. Plant pattern development during early post-
restoration succession in grasslands-a case study of *Arabis nemorensis*. *Restoration Ecology* **19**.

Cai Z, Zhou L, Ren NN, Xu X, Liu R, Huang L, Zheng XM, Meng QL, Du YS, Wang MX, *et al.* 2019. Parallel
Speciation of Wild Rice Associated with Habitat Shifts. *Molecular Biology and Evolution* **36**.

Carey K, Ganders FR. **1980**. Heterozygote Advantage at the Fruit Wing Locus in *Plectritis congesta*
(Valerianaceae). *Evolution* **34**.

Catchen J, Hohenlohe PA, Bassham S, Amores A, Cresko WA. **2013**. Stacks: An analysis tool set for
population genomics. *Molecular Ecology* **22**.

Ceballos G, Ehrlich PR, Raven PH. **2020**. Vertebrates on the brink as indicators of biological
annihilation and the sixth mass extinction. *Proceedings of the National Academy of Sciences* **117**:
13596–13602.

Cerise M, da Silveira Falavigna V, Rodríguez-Maroto G, Signol A, Severing E, Gao H, van Driel A,
Vincent C, Wilkens S, Iacobini FR, *et al.* 2023. Two modes of gene regulation by *TFL1* mediate its dual
function in flowering time and shoot determinacy of *Arabidopsis*. *Development (Cambridge)* **150**.

Cheng H, Concepcion GT, Feng X, Zhang H, Li H. **2021**. Haplotype-resolved de novo assembly using
phased assembly graphs with hifiasm. *Nature Methods* **18**.

Cheng H, Jarvis ED, Fedrigo O, Koepfli KP, Urban L, Gemmell NJ, Li H. **2022**. Haplotype-resolved
assembly of diploid genomes without parental data. *Nature Biotechnology*.

Chettoor AM, Givan SA, Cole RA, Coker CT, Unger-Wallace E, Vejlupkova Z, Vollbrecht E, Fowler JE,
Evans MMS. **2014**. Discovery of novel transcripts and gametophytic functions via RNA-seq analysis of
maize gametophytic transcriptomes. *Genome Biology* **15**.

Chunco AJ. **2014**. Hybridization in a warmer world. *Ecology and Evolution* **4**.

Clarke HJ, Khan TN, Siddique KHM. **2004**. Pollen selection for chilling tolerance at hybridisation leads
to improved chickpea cultivars. *Euphytica* **139**.

Clo J, Ronfort J, Gay L. **2021**. Fitness consequences of hybridization in a predominantly selfing species:
insights into the role of dominance and epistatic incompatibilities. *Heredity* **127**.

Cochrane SKJ, Andersen JH, Berg T, Blanchet H, Borja A, Carstensen J, Elliott M, Hummel H, Niquil N,
Renaud PE. **2016**. What is marine biodiversity? Towards common concepts and their implications for
assessing biodiversity status. *Frontiers in Marine Science* **3**.

Colloff MJ, Lavorel S, Wise RM, Dunlop M, Overton IC, Williams KJ. **2016**. Adaptation services of
floodplains and wetlands under transformational climate change. *Ecological Applications* **26**.

Cooper BS, Sedghifar A, Nash WT, Comeault AA, Matute DR. **2018**. A Maladaptive Combination of
Traits Contributes to the Maintenance of a *Drosophila* Hybrid Zone. *Current Biology* **28**.

[Core Writing Team HL and JR (eds.). **2022**. Climate Change 2023: Synthesis Report. Contribution of
Working Groups I, II and III to the Sixth Assessment Report of the Intergovernmental Panel on Climate
Change. *IPCC* **13**.

Le Corre V, Roux F, Reboud X. **2002**. DNA polymorphism at the *FRIGIDA* gene in *Arabidopsis thaliana*:
Extensive nonsynonymous variation is consistent with local selection for flowering time. *Molecular*
Biology and Evolution **19**.

Coughlan JM, Matute DR. **2020**. The importance of intrinsic postzygotic barriers throughout the
speciation process: Intrinsic barriers throughout speciation. *Philosophical Transactions of the Royal*
Society B: Biological Sciences **375**.

Cowie RH, Bouchet P, Fontaine B. **2022**. The Sixth Mass Extinction: fact, fiction or speculation?
Biological Reviews **97**.

Danecek P, Auton A, Abecasis G, Albers CA, Banks E, DePristo MA, Handsaker RE, Lunter G, Marth
GT, Sherry ST, *et al.* 2011. The variant call format and VCFtools. *Bioinformatics* **27**.

Delaneau O, Marchini J, McVeanh GA, Donnelly P, Lunter G, Myers S, Gupta-Hinch A, Iqbal Z,
Mathieson I, Rimmer A, *et al.* 2014. Integrating sequence and array data to create an improved 1000
Genomes Project haplotype reference panel. *Nature Communications* **5**.

Delaneau O, Ongen H, Brown AA, Fort A, Panousis NI, Dermitzakis ET. **2017**. A complete tool set for
molecular QTL discovery and analysis. *Nature Communications* **8**.

Dittberner H, Becker C, Jiao WB, Schneeberger K, Hölzel N, Tellier A, de Meaux J. **2019**. Strengths
and potential pitfalls of hay transfer for ecological restoration revealed by RAD-seq analysis in
floodplain *Arabis* species. *Molecular Ecology* **28**.

Dittberner H, Tellier A, De Meaux J. **2022**. Approximate Bayesian Computation Untangles Signatures
of Contemporary and Historical Hybridization between Two Endangered Species. *Molecular Biology*
and Evolution **39**.

Dobzhansky T. **1936**. STUDIES ON HYBRID STERILITY. II. LOCALIZATION OF STERILITY FACTORS IN
DROSOPHILA PSEUDOOBSCURA HYBRIDS. *Genetics* **21**.

Domínguez E, Cuartero J, Fernández-Muñoz R. **2005**. Breeding tomato for pollen tolerance to low
temperatures by gametophytic selection. *Euphytica* **142**.

Durand NC, Robinson JT, Shamim MS, Machol I, Mesirov JP, Lander ES, Aiden EL. 2016a. *Cell Systems*
3: 99–101.

Durand NC, Shamim MS, Machol I, Rao SS, Huntley MH, Lander ES, Aiden EL. 2016b. *Cell Systems* **3:**
95–98.

Eichenberg D, Bowler DE, Bonn A, Bruelheide H, Grescho V, Harter D, Jandt U, May R, Winter M,
Jansen F. **2021**. Widespread decline in Central European plant diversity across six decades. *Global*
Change Biology **27**.

Ellstrand NC, Schierenbeck KA. **2000**. Hybridization as a stimulus for the evolution of invasiveness in plants? *Proceedings of the National Academy of Sciences of the United States of America* **97**.

Finseth F. **2023**. Female meiotic drive in plants: mechanisms and dynamics. *Current Opinion in Genetics and Development* **82**.

Finseth FR, Dong Y, Saunders A, Fishman L. **2015**. Duplication and adaptive evolution of a key centromeric protein in mimulus, a genus with female meiotic drive. *Molecular Biology and Evolution* **32**.

Fishman L, Kelly JK. **2015**. Centromere-associated meiotic drive and female fitness variation in *Mimulus*. *Evolution* **69**.

Frankham R. **2015**. Genetic rescue of small inbred populations: meta-analysis reveals large and consistent benefits of gene flow. *Molecular Ecology* **24**.

Goulet BE, Roda F, Hopkins R. **2017**. Hybridization in plants: Old ideas, new techniques[OPEN]. *Plant Physiology* **173**.

Haley CS, Knott SA. **1992**. A simple regression method for mapping quantitative trait loci in line crosses using flanking markers. *Heredity* **69**.

Haller BC, Galloway J, Kelleher J, Messer PW, Ralph PL. **2019**. Tree-sequence recording in SLiM opens new horizons for forward-time simulation of whole genomes. *Molecular Ecology Resources* **19**.

Haller BC, Messer PW. **2023**. SLiM 4: Multispecies Eco-Evolutionary Modeling. *American Naturalist* **201**.

Hancock A, Portalier S, Fulgione A, Stetter MG, de Meaux J. **2024**. The molecular basis of adaptation to climatic factors and range change in plants. Annual reviews in Ecology and Evolution. *Submitted*.

Hansen MM, Olivieri I, Waller DM, Nielsen EE. **2012**. Monitoring adaptive genetic responses to environmental change. *Molecular Ecology* **21**.

Hölzel N. **2005**. Seedling recruitment in flood-meadow species: The effects of gaps, litter and vegetation matrix. *Applied Vegetation Science* **8**.

Hooper JNA, Kennedy JA, Quinn RJ. **2002**. Biodiversity 'hotspots', patterns of richness and endemism, and taxonomic affinities of tropical Australian sponges (Porifera). *Biodiversity and Conservation* **11**.

Hopkins R. **2013**. Reinforcement in plants. *New Phytologist* **197**.

Karl R, Koch MA. **2014**. Phylogenetic signatures of adaptation: The *Arabis hirsuta* species aggregate (Brassicaceae) revisited. *Perspectives in Plant Ecology, Evolution and Systematics* **16**.

Klepikova A V., Kasianov AS, Gerasimov ES, Logacheva MD, Penin AA. **2016**. A high resolution map of the *Arabidopsis thaliana* developmental transcriptome based on RNA-seq profiling. *Plant Journal* **88**.

Kraft TS, Wright SJ, Turner I, Lucas PW, Oufiero CE, Supardi Noor MN, Sun IF, Dominy NJ. **2015**. Seed size and the evolution of leaf defences. *Journal of Ecology* **103**.

Lankinen S, Maad J, Armbruster WS. **2009**. Pollen-tube growth rates in *Collinsia heterophylla* (Plantaginaceae): One-donor crosses reveal heritability but no effect on sporophytic-offspring fitness. *Annals of Botany* **103**.

Lewontin RC, Birch LC. **1966**. Hybridization as a Source of Variation for Adaptation to New
Environments. *Evolution* **20**.

Li H, Durbin R. **2009**. Fast and accurate short read alignment with Burrows-Wheeler transform.
Bioinformatics **25**.

Li H, Handsaker B, Wysoker A, Fennell T, Ruan J, Homer N, Marth G, Abecasis G, Durbin R. **2009**. The
Sequence Alignment/Map format and SAMtools. *Bioinformatics* **25**.

Li J, Schumer M, Bank C. **2022**. Imbalanced segregation of recombinant haplotypes in hybrid
populations reveals inter- and intrachromosomal Dobzhansky-Muller incompatibilities. *PLoS Genetics*
18.

Lundgren MR, Des Marais DL. **2020**. Life History Variation as a Model for Understanding Trade-Offs in
Plant–Environment Interactions. *Current Biology* **30**.

Ma Y, Marczewski T, Xue D, Wu Z, Liao R, Sun W, Marczewski J. **2019**. Conservation implications of
asymmetric introgression and reproductive barriers in a rare primrose species. *BMC Plant Biology* **19**.

Malik HS. **2005**. *Mimulus* finds centromeres in the driver's seat. *Trends in Ecology and Evolution* **20**.

Mallet J. **2007**. Hybrid speciation. *Nature* **446**: 279–283.

Marçais G, Kingsford C. **2011**. A fast, lock-free approach for efficient parallel counting of occurrences
of k-mers. *Bioinformatics* **27**: 764–770.

Martin M. **2011**. Cutadapt removes adapter sequences from high-throughput sequencing reads.
EMBnet.journal **17**.

Masly JP, Presgraves DC. **2007**. High-resolution genome-wide dissection of the two rules of speciation
in *Drosophila*. *PLoS Biology* **5**.

Mathar W, Kleinebecker T, Hölzel N. **2015**. Environmental variation as a key process of co-existence
in flood-meadows. *Journal of Vegetation Science* **26**.

McKenna A, Hanna M, Banks E, Sivachenko A, Cibulskis K, Kernytsky A, Garimella K, Altshuler D,
Gabriel S, Daly M, *et al.* 2010. The genome analysis toolkit: A MapReduce framework for analyzing
next-generation DNA sequencing data. *Genome Research* **20**.

Moore JS, Hendry AP. **2009**. Can gene flow have negative demographic consequences? Mixed
evidence from stream threespine stickleback. *Philosophical Transactions of the Royal Society B:*
Biological Sciences **364**.

Moraes TS, Dornelas MC, Martinelli AP. **2019**. *FT/TFL1*: Calibrating plant architecture. *Frontiers in*
Plant Science **10**.

Moran BM, Payne CY, Powell DL, Iverson ENK, Donny AE, Banerjee SM, Langdon QK, Gunn TR,
Rodriguez-Soto RA, Madero A, *et al.* 2024. A lethal mitonuclear incompatibility in complex I of natural
hybrids. *Nature* **626**.

Muller HJ. **1942**. Isolating mechanisms, evolution and temperature. *Biol. Symp* **6**.

Nocchi G, Wang J, Yang L, Ding J, Gao Y, Buggs RJA, Wang N. **2023**. Genomic signals of local adaptation
and hybridization in Asian white birch. *Molecular Ecology* **32**.

Okonechnikov K, Conesa A, García-Alcalde F. **2016**. Qualimap 2: Advanced multi-sample quality
control for high-throughput sequencing data. *Bioinformatics* **32**.

Parker MT, Amar S, Campoy JA, Krause K, Tusso S, Marek M, Huettel B, Schneeberger K. **2025**.
Scalable eQTL mapping using single-nucleus RNA-sequencing of recombined gametes from a small
number of individuals. *PLOS Biology* **23**.

Peñalba J V., Runemark A, Meier JI, Singh P, Wogan GOU, Sánchez-Guillén R, Mallet J, Rometsch SJ,
Menon M, Seehausen O, *et al.* 2024. The Role of Hybridization in Species Formation and Persistence.
Cold Spring Harbor Perspectives in Biology.

Pfennig KS, Kelly AL, Pierce AA. **2016**. Hybridization as a facilitator of species range expansion.
Proceedings of the Royal Society B: Biological Sciences **283**.

Pinkas R, Zamir D, Ladizinsky G. **1985**. Allozyme divergence and evolution in the genusLens. *Plant*
Systematics and Evolution **151**: 131–140.

Powell DL, García-Olazábal M, Keegan M, Reilly P, Du K, Díaz-Loyo AP, Banerjee S, Blakkan D, Reich
D, Andolfatto P, *et al.* 2020. Natural hybridization reveals incompatible alleles that cause melanoma
in swordtail fish. *Science* **368**.

Presgraves DC. **2010**. The molecular evolutionary basis of species formation. *Nature Reviews Genetics*
11.

Ranallo-Benavidez TR, Jaron KS, Schatz MC. **2020**. GenomeScope 2.0 and Smudgeplot for reference-
free profiling of polyploid genomes. *Nature Communications* **11**.

Rhymer JM, Simberloff D. **1996**. Extinction by hybridization and introgression. *Annual Review of*
Ecology and Systematics **27**.

Rieseberg LH, Raymond O, Rosenthal DM, Lai Z, Livingstone K, Nakazato T, Durphy JL, Schwarzbach
AE, Donovan LA, Lexer C. **2003**. Major ecological transitions in wild sunflowers facilitated by
hybridization. *Science* **301**.

Rivera-Colón AG, Catchen J. **2022**. Population Genomics Analysis with RAD, Reprised: Stacks 2. In:
Methods in Molecular Biology.

Rosser N, Seixas F, Queste LM, Cama B, Mori-Pezo R, Kryvokhyzha D, Nelson M, Waite-Hudson R,
Goringe M, Costa M, *et al.* 2024. Hybrid speciation driven by multilocus introgression of ecological
traits. *Nature* **628**: 811–817.

Rutley N, Twell D. **2015**. A decade of pollen transcriptomics. *Plant Reproduction* **28**.

Sandler L, Novitski E. **1957**. Meiotic Drive as an Evolutionary Force. *The American Naturalist* **91**.

Schlaepfer MA, Lawler JJ. **2023**. Conserving biodiversity in the face of rapid climate change requires a
shift in priorities. *Wiley Interdisciplinary Reviews: Climate Change* **14**.

Schnittler M, Günther KF. **1999**. Central European vascular plants requiring priority conservation
measures - An analysis from national Red Lists and distribution maps. *Biodiversity and Conservation* **8**.

Schumer M, Brandvain Y. **2016**. Determining epistatic selection in admixed populations. *Molecular*
ecology **25**.

Schumer M, Cui R, Powell DL, Dresner R, Rosenthal GG, Andolfatto P. **2014**. High-resolution mapping reveals hundreds of genetic incompatibilities in hybridizing fish species. *eLife* **2014**.

Schumer M, Cui R, Rosenthal GG, Andolfatto P. **2015**. Reproductive Isolation of Hybrid Populations Driven by Genetic Incompatibilities. *PLoS Genetics* **11**.

Schumer M, Xu C, Powell DL, Durvasula A, Skov L, Holland C, Blazier JC, Sankararaman S, Andolfatto P, Rosenthal GG, *et al.* **2018**. Natural selection interacts with recombination to shape the evolution of hybrid genomes. *Science* **360**.

Seehausen O. **2004**. Hybridization and adaptive radiation. *Trends in Ecology and Evolution* **19**.

Siegmund JF, Wiedermann M, Donges JF, Donner R V. **2016**. Impact of temperature and precipitation extremes on the flowering dates of four German wildlife shrub species. *Biogeosciences* **13**.

Sim SB, Corpuz RL, Simmonds TJ, Geib SM. **2022**. HiFiAdapterFilt, a memory efficient read processing pipeline, prevents occurrence of adapter sequence in PacBio HiFi reads and their negative impacts on genome assembly. *BMC Genomics* **23**.

Simons YB, Sella G. **2016**. The impact of recent population history on the deleterious mutation load in humans and close evolutionary relatives. *Current Opinion in Genetics and Development* **41**.

Smit AFA, Hubley R, Grenn P. **2015**. RepeatMasker Open-4.0. *RepeatMasker Open-4.0.7.*

Smith LM, Bomblies K, Weigel D. **2011**. Complex evolutionary events at a tandem cluster of *Arabidopsis thaliana* genes resulting in a single-locus genetic incompatibility. *PLoS Genetics* **7**.

Snow AA, Spira TP. **1996**. Pollen-tube competition and male fitness in *Hibiscus moscheutos*. *Evolution* **50**.

Staudinger MD, Grimm NB, Staudt A, Carter SF, Chapin FS, Kareiva P, Ruckelshaus M, Stein BA. **2012**. Impacts of climate change on biodiversity, ecosystems, and ecosystem services: technical input to the 2013 National Climate Assessment. *Cooperative Report to the 2013 National Climate Assessment*.

Stull GW, Pham KK, Soltis PS, Soltis DE. **2023**. Deep reticulation: the long legacy of hybridization in vascular plant evolution. *Plant Journal* **114**.

Taylor J, Butler D. **2017**. R package ASMap: Efficient genetic linkage map construction and diagnosis. *Journal of Statistical Software* **79**.

Theissinger K, Fernandes C, Formenti G, Bista I, Berg PR, Bleidorn C, Bombarely A, Crottini A, Gallo GR, Godoy JA, *et al.* **2023**. How genomics can help biodiversity conservation. *Trends in Genetics* **39**.

Titz W. **1979**. Die Interfertilitätsbeziehungen europäischer Sippen der *Arabis hirsuta*-Gruppe (Brassicaceae). *Plant Systematics and Evolution* **131**.

Todesco M, Owens GL, Bercovich N, Légaré JS, Soudi S, Burge DO, Huang K, Ostevik KL, Drummond EBM, Imerovski I, *et al.* **2020**. Massive haplotypes underlie ecotypic differentiation in sunflowers. *Nature* **584**.

Todesco M, Pascual MA, Owens GL, Ostevik KL, Moyers BT, Hübner S, Heredia SM, Hahn MA, Caseys C, Bock DG, *et al.* **2016**. Hybridization and extinction. *Evolutionary Applications* **9**.

Tun W, Yoon J, Jeon JS, An G. **2021**. Influence of Climate Change on Flowering Time. *Journal of Plant Biology* **64**.

Villasenor Alva JA, Estrada EG. **2009**. A generalization of Shapiro-Wilk's test for multivariate normality.
Communications in Statistics - Theory and Methods **38**.

Wang R, Albani MC, Vincent C, Bergonzi S, Luan M, Bai Y, Kiefer C, Castillo R, Coupland G. **2011**. Aa
tfl1 confers an age-dependent response to vernalization in perennial *Arabis alpina*. *Plant Cell* **23**.

Wang Z, Jiang Y, Bi H, Lu Z, Ma Y, Yang X, Chen N, Tian B, Liu B, Mao X, *et al.* 2021. Hybrid speciation
via inheritance of alternate alleles of parental isolating genes. *Molecular Plant* **14**.

Wang RJ, White MA, Payseur BA. **2015**. The Pace of hybrid incompatibility evolution in house mice.
Genetics **201**.

White MA, Steffy B, Wiltshire T, Payseur BA. **2011**. Genetic dissection of a key reproductive barrier
between nascent species of house mice. *Genetics* **189**.

White MA, Stubbings M, Dumont BL, Payseur BA. **2012**. Genetics and evolution of hybrid male
sterility in house mice. *Genetics* **191**.

Wickham H. **2011**. ggplot2. *Wiley Interdisciplinary Reviews: Computational Statistics* **3**: 180–185.

Zuellig MP, Sweigart AL. **2018**. Gene duplicates cause hybrid lethality between sympatric species of
Mimulus. *PLoS Genetics* **14**.

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