Tracking genes of ecological relevance using a genome scan in two independent regional population samples of Arabis alpina

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Abstract

Understanding the genetic basis of adaptation in response to environmental variation is fundamental as adaptation plays a key role in the extension of ecological niches to marginal habitats and in ecological speciation. Based on the assumption that some genomic markers are correlated to environmental variables, we aimed to detect loci of ecological relevance in the alpine plant Arabis alpina L. sampled in two regions, the French (99 locations) and the Swiss (109 locations) Alps. We used an unusually large genome scan [825 amplified fragment length polymorphism loci (AFLPs)] and four environmental variables related to temperature, precipitation and topography. We detected linkage disequilibrium among only 3.5% of the considered AFLP loci. A population structure analysis identified no admixture in the study regions, and the French and Swiss Alps were differentiated and therefore could be considered as two independent regions. We applied generalized estimating equations (GEE) to detect ecologically relevant loci separately in the French and Swiss Alps. We identified 78 loci of ecological relevance (9%), which were mainly related to mean annual minimum temperature. Only four of these loci were common across the French and Swiss Alps. Finally, we discuss that the genomic characterization of these ecologically relevant loci, as identified in this study, opens up new perspectives for studying functional ecology in A. alpina, its relatives and other alpine plant species.

Keywords: adaptive genetic variation, amplified fragment length polymorphism, generalized estimating equations, landscape genomics, local adaptation

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Introduction

Adaptation is the evolution of a plant population towards a phenotype that better fits the present environment (Orr 2005). The major driving force of adaptation is divergent selection (Kawecki 2008). Understanding the genetic changes resulting in divergent selection is a major objective in evolutionary genetics (Feder & Mitchell-Olds 2003), since adaptation plays a key role in the extension of the ecological niche to marginal habitats (Kawecki 2008) as well as in ecological speciation (Schluter & Conte 2009). Adaptation to new habitats is also a potential plant response to shifts in environmental conditions, which is all the more crucial in the context of climate change (Reusch & Wood 2007). Until recently, it has been difficult to detect genetic variation of adaptive relevance.
in nonmodel organisms (Bonin et al. 2006). Genome scans combined with landscape genomic approaches have the potential to improve our understanding of the genetic basis of adaptation in nonmodel species with unknown genomes (Storz 2005) through the investigation of a large number of loci distributed across a genome in a large number of individuals (Black et al. 2001; Luikart et al. 2003; Schlötterer 2003). In nonmodel plants, genome scans have been used to detect ecologically relevant loci (ERLs) under various conditions (Scotti-Saintagne et al. 2004; Savolainen et al. 2006; Parisod & Christin 2008). However, underlying environmental cues or even functional traits involved in adaptation have rarely been identified in these studies (Vasemägi & Primmer 2005). For example, Namroud et al. (2008) tracked candidate genes for local adaptation in six Canadian populations of white spruce (Picea glauca) using a genome scan of expressed sequence tags (EST). Amongst the genes identified as potentially being under natural selection, half were specific to populations in the warmest locations and half to populations experiencing different extreme conditions (most arid, coldest or most humid locations).

Allele distribution models (Holderegger et al. 2007), which relate allele frequencies to environmental variables, aim to identify genes of ecological relevance. However, patterns of genetic variation that seem to be related to natural selection may also result from historical and spatial effects (Endler 1986; Kawecki & Ebert 2004; Schmidt et al. 2008). Firstly, isolation by distance might restrict gene flow among populations. Thereby, neutral alleles will change in frequency due to genetic drift (Wright 1938). However, this trend would affect the entire genome and not only specific loci. Secondly, contact and admixture zones, where populations which have diverged in isolation (e.g. in different glacial refugia) come into secondary contact, may result in confusing genetic patterns due to past selection followed by a spatial gradient of population admixture (Endler 1977). Using replicated regions and searching for congruent patterns of ERLs across regions is one way of limiting the confounding influence of historical and spatial processes, as it is unlikely that these neutral processes and genetic drift would generate similar genetic patterns at a locus across independent environmental gradients (e.g. see Wark & Peichl 2010).

The objective of this study was to identify loci of ecological relevance in the perennial alpine herb Arabis alpina. We did so based on the premise that loci whose distributions are correlated to environmental variation are ecologically relevant and could be linked to genomic regions undergoing selection. Although loci of ecological relevance might well be region-specific, more weight will be given to loci identified in two independent regions. To investigate the above assumption, we developed a large data set of more than 800 amplified fragment length polymorphism (AFLP) loci, greatly exceeding the numbers usually tested in genome scan studies. Prior to the outlier analysis, we checked for linkage disequilibrium in the set of loci and elucidated historical and spatial effects by evaluating population structure, testing an admixture model and investigating isolation by distance. To correlate allele frequencies to environmental variables, we used generalized estimating equations (GEE) with binomial error. The GEE method makes it possible to consider the spatial autocorrelation of individuals within sampling locations (Carl & Kuhn 2007), which can be high in the case of sessile plants. Finally, the ERLs identified are discussed with regard to available genomic resources (i.e. the Arabidopsis thaliana whole genome sequence), opening up new perspectives for studies in functional ecology.

Materials and methods

Study species and area

Arabis alpina L. (Brassicaceae) is a perennial arctic-alpine rosette plant widely distributed over the Arctic and mountainous regions of the northern hemisphere. It is particularly common in the European Alps (Bovet et al. 2006), where it shows a large altitudinal distribution, ranging from mountain forests to alpine and even nival vegetation patches. A. alpina prefers moist, open and rocky habitats, and reproduces sexually or asexually by stoloniferous growth (Ansell et al. 2008). Its small seeds are dispersed by wind.

Plants were sampled from 99 locations in the French Alps and from 109 locations in the Swiss Alps (Table S1, Supporting information). In the French Alps, A. alpina was mainly sampled in three massifs: Vercors, Chartreuse and Briançonnais. In the Swiss Alps, sampling locations were distributed across south-eastern Switzerland (Fig. 1). They were chosen to cover a wide range of habitats (e.g. screes, moist, nutrient rich) and elevations (440–3133 m above sea level). Geographical distances between locations ranged from 23 m to 143 km in the French Alps and from 164 m to 132 km in the Swiss Alps. Fresh plant material from nine individuals per location (in the vast majority of cases) was collected between July and September 2006 and immediately dried in silica gel.

Environmental variables

A set of 20 monthly and annual environmental variables related to temperature, precipitation and topography was extracted for each sampling location from published GIS eco-climatic layers at the spatial resolution of 200 m
A principal component analysis was used to check for the correlation between these eco-climatic variables and elevation. Only the four most uncorrelated explanatory variables were selected from PCA: mean annual minimum temperature (tmin), sum of precipitation between March and May (prcp), slope index (slp) and topographic wetness index (twi; i.e. the ratio of the upslope contributing region on the tangent of the slope angle). The variable twi is a hydrological index developed by Beven & Kirkby (1979) as a measure of soil moisture at a spatial resolution of 400 m. These four selected variables showed variation among the sites sampled in the two study regions. They were used as explanatory environmental variables in the identification of loci considered to have undergone divergent selection.

**DNA extraction and AFLP genotyping**

Total DNA from three, and in few cases nine individuals per sampling location was extracted using the DNeasy 96 plant extraction kit (QIAGEN) according to the manufacturer’s instructions. We used AFLPs (Vos et al. 1995), which have become the reference marker technique for genome scans (Bonin et al. 2007; Meudt & Clarke 2007), because AFLP loci are widely distributed across the genome. AFLP loci were generated for 768 samples. The AFLP procedure followed Vos et al. (1995) with minor modifications. Digestion of genomic DNA was performed for 2 h at 37 °C in a 20 µL mix using 2 U of MseI and 5 U of EcoRI or PstI (New England Biolabs). Double-stranded adaptors were then ligated to the digested DNA in a 40 µL volume for 2 h at 37 °C using 1 U of T4 DNA Ligase (Roche). Products were diluted 1:9, and pre-selective PCR (120 s at 94 °C, 30 cycles of 30 s at 94 °C, 30 s at 56 °C, 120 s at 72 °C with a final elongation of 10 min at 72 °C) was carried out in a 25 µL volume containing 1X PCR Buffer II pH 8.3, 1.5 mM of MgCl2, 80 µM of each dNTP, 0.2 µM of each primer at 10 µM (EcoRI/MseI or PstI/MseI), and 0.5 U of AmpliTaq DNA polymerase (Applied Biosystems). After 1:19 dilution of pre-selective PCR products, the selective amplification (10 min at 95 °C, 13 cycles of 30 s at 94 °C, 60 s at 65 °C to 56 °C, and 60 s at 72 °C, 23 cycles of 30 s at 94 °C, 60 s at 56 °C, and 60 s at 72 °C with a final elongation of 10 min at 72 °C) was carried out in a 12.5 µL volume containing 2.5 µL of diluted, FAM-labelled products were mixed with 10 µL of HiDi formamide and 0.1 µL Genescan ROX 500 size standard (Applied Biosystems) and electrophoresed on an ABI 3100. The PCR products were sized using a GeneScan 500 ROX size standard (Applied Biosystems). After preliminary tests, 19 primer combinations were chosen which resulted in clear bands of sufficient variability (Table 1). PCR products of EcoRI/MseI combinations were purified using columns of half to half 5% Sephadex G50 and Sephadryl S200, diluted 1:9. PCR products from each EcoRI/MseI primer pair were run separately, but multiplexed for PstI/MseI primer pairs for fragment length analysis. For EcoRI/MseI combinations, 1.5 µL of the diluted, FAM-labelled products were mixed with 10 µL of HiDi formamide and 0.1 µL Genescan ROX 500 size standard (Applied Biosystems) and electrophoresed on an ABI 3100.
Table 1 Number of DNA fragments generated by 19 AFLP primer–enzyme combinations used in Arabis alpina

<table>
<thead>
<tr>
<th>Primer–enzyme combination</th>
<th>Number of loci</th>
<th>Percentage of polymorphic loci (%)</th>
<th>Mean error rate (%)</th>
</tr>
</thead>
<tbody>
<tr>
<td>EcoRI-AAT/Msel-CAC</td>
<td>100</td>
<td>37.0</td>
<td>1.2</td>
</tr>
<tr>
<td>EcoRI-AGC/Msel-CAC</td>
<td>135</td>
<td>68.1</td>
<td>0.8</td>
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<tr>
<td>EcoRI-ATC/Msel-CAC</td>
<td>126</td>
<td>48.4</td>
<td>0.7</td>
</tr>
<tr>
<td>EcoRI-AGG/Msel-CAC</td>
<td>90</td>
<td>43.3</td>
<td>1.2</td>
</tr>
<tr>
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<td>102</td>
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<td>1.4</td>
</tr>
<tr>
<td>EcoRI-AGC/Msel-CTG</td>
<td>112</td>
<td>43.8</td>
<td>0.8</td>
</tr>
<tr>
<td>EcoRI-AGC/Msel-CTC</td>
<td>106</td>
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<td>1.2</td>
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<tr>
<td>PstI-AAG/Msel-CA</td>
<td>108</td>
<td>40.7</td>
<td>1.3</td>
</tr>
<tr>
<td>PstI-ACT/Msel-CA</td>
<td>84</td>
<td>50.0</td>
<td>1.0</td>
</tr>
<tr>
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<td>63</td>
<td>39.7</td>
<td>1.2</td>
</tr>
<tr>
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<td>96</td>
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<td>1.5</td>
</tr>
<tr>
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<td>1.0</td>
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<tr>
<td>PstI-AGA/Msel-CT</td>
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<td>34.7</td>
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<td>1.3</td>
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<tr>
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<td>56.2</td>
<td>1.2</td>
</tr>
<tr>
<td>PstI-AGA/Msel-CT</td>
<td>66</td>
<td>50.0</td>
<td>1.9</td>
</tr>
<tr>
<td>PstI-AAC/Msel-CT</td>
<td>48</td>
<td>50.0</td>
<td>2.3</td>
</tr>
<tr>
<td>Total</td>
<td>1731</td>
<td>43.4</td>
<td>47.3</td>
</tr>
<tr>
<td>Mean</td>
<td>91.1</td>
<td>43.4</td>
<td>1.2</td>
</tr>
</tbody>
</table>

*Only selective bases are given for each enzyme-specific primer.
†A locus was considered polymorphic if at least three individuals showed a different allele than all other samples.
‡Mean error rates were calculated by summing differences between control profiles (i.e. samples that were genotyped twice) per primer–enzyme combination.

PRISM 3100 capillary sequencer (Applied Biosystems). For PstI/Msel combinations, 1 μL of each of the diluted PCR products was mixed with 10 μL of HiDi formamide and 0.15 μL Genescan ROX 500 size standard (Applied Biosystems) and electrophoresed on an ABI PRISM 3100-Avant capillary sequencer (Applied Biosystems). Raw data were sized with GeneMapper 3.7 (Applied Biosystems).

AFLP loci were obtained using a semi-automated procedure. Loci were defined manually in GeneMapper and scored for the presence or absence of AFLP fragments in each individual. The quality of all loci was then automatically checked with an R script as described in Herrmann et al. (2010; http://www-leca.ujf-grenoble.fr/logic/els.htm). This script assesses presence/absence within a particular locus based on peak height distribution and further removes loci with large peak height variation or low reproducibility. AFLP fragments shorter than 50 bp for EcoRI/Msel or 75 bp for PstI/Msel primer–enzyme combinations were discarded. We included 40 control samples from DNA extraction throughout AFLP analysis to check for locus reliability (replicates between and within plates). In addition, we analysed 39 duplicated samples as blind controls. The mismatch error rate per primer–enzyme combination was calculated as the percentage of loci with different scores between these blind controls and their respective original samples after locus selection and phenotype identification (Bonin et al. 2004; Pompanon et al. 2005; Herrmann et al. 2010; Table 1). Loci were considered as polymorphic when at least three individuals had a different score than all other individuals.

Linkage disequilibrium among AFLP loci was tested using successive logistic regressions between each locus as the response variable and the others as explanatory variables. We also applied a principal component analysis to investigate correlations among the occurrences of AFLP loci within sites. These analyses were performed using the R software (R Development Core Team, 2007).

Population structure, historical and spatial effects

Here, our main objective was to investigate whether the French and the Swiss Alps could be considered as two independent study regions with no admixture due to secondary contact. We conducted an individual-level analysis to check the population structure of our dataset using STRUCTURE 2.2 adapted for dominant markers (Falush et al. 2007) with and without an admixture model. We determined the optimal model by testing a range of values for the number of populations (K). For each value of K, seven runs were conducted. The number of populations K was approximated as the highest estimated log probability of the data Pr(X|K) (Pritchard et al. 2000). Because Pr(X|K) often reaches a plateau for maximal K values and to be more discriminatory, we also used the approach introduced by Evanno et al. (2005) based on the computation of the ΔK. ΔK is defined as (1/nL(K+1) − 2mL(K) + mL(K − 1))/sd1(K) with mL and sd1 being the mean and the standard deviation of the likelihood value across the runs. As recommended by these authors, we used the number of populations of the modal value of the ΔK distribution as the correct estimation of K. The STRUCTURE 2.2 runs were conducted for each value of K ranging between 1 and 8 using 50 000 burn in periods followed by 50 000 Markov chain Monte Carlo (MCMC) repetitions. In both the admixture and the nonadmixture models, the correlated allele frequency option was used.

Genetic differentiation between the two sampling regions was estimated with AFLP-SURV (Vekemans et al. 2002) using the Bayesian method with nonuniform prior distribution and assuming Hardy–Weinberg equilibrium (FIS = 0). Although this last assumption is
expected to be violated (preliminary not shown results), Ansell et al. (2008) and Bonin et al. (2007) have shown that multilocus estimates of \( F_{ST} \) were robust to change in assumed \( F_{IS} \). Genetic differentiation between sampling locations was estimated (1000 permutations).

To test for possible spatial effects, we calculated isolation by distance (IBD) estimates. For each region, pairwise kinship coefficients \( F_{ij} \) for all sample pairs, adjusted for dominant loci and assuming no inbreeding (Hardy 2003), were estimated using SPAGeDi (Hardy & Vekemans 2002). Kinship under IBD is expected to decrease in line with geographic distance. The correlations between pairwise \( F_{ij} \) and both untransformed and log-transformed Euclidian distances were inferred (999 permutations) and plotted for 1000 classes of geographic distances.

**Detection of ecologically relevant loci using GEE**

To detect alleles correlated to environmental variables, we used generalized estimating equations (GEE), which are an extension of generalized linear models (Diggle & Elliott 1995; Carl & Kuhn 2007). GEE consider spatial autocorrelation between individuals collected at the same sampling location by including an additional variance component to accommodate correlated data. Individuals collected in the same location were considered to belong to the same cluster in the model. Therefore, each cluster usually consisted of three (rarely nine) plants, with 99 clusters in one region (French Alps) and 109 in the other region (Swiss Alps). This method makes it possible to take into account the fact that neighbouring individuals within locations (i.e. at distances of only some metres) are genetically more similar than distant individuals between locations (inter-location distances: 23–163 km). As we dealt with binary data, we used a logit-link and binomial error distribution to correlate allele occurrence for each AFLP locus per sampling location to quadratic polynomials of environmental variables. Using quadratic polynomials makes it possible to account for the variety of response curve shapes other than a linear response (Legendre & Legendre 1998). GEE are nonlikelihood-based models and cannot be compared using a maximum likelihood methods such as Akaike’s information criterion (AIC). Instead, we used the quasi-likelihood information criterion (QIC) adapted for the selection of GEE models by Pan (2001). The best model is that with the lowest QIC. All combinations of variables were investigated for each locus. Finally, we tested 255 GEE (from the null model to the full model with four variables in second-order polynomials) separately in the French data set and in the Swiss data set. All GEE models were calculated using the R package geepack (Yan & Fine 2004), and we implemented the QIC calculation in R for this study (R Development Core Team 2007).

For each selected model, we also tested each regression coefficient using the Wald test. To avoid a high rate of false positives due to multiple tests, significance levels were calculated for each regional data set by minimizing the false discovery rate (FDR; Storey 2002). \( P \)-value rejection thresholds were adjusted so as to have less than one false positive locus per regional data set and were estimated using the R package qvalue.

**Results**

**Locus polymorphisms**

A total of 679 individuals from 99 sampling locations in the French Alps and 109 in the Swiss Alps (1 to 9 individuals per location; Table S1, Supporting information) were reliably scored for the presence or absence of AFLP fragments. Overall, 1731 reliable loci were identified from the 19 selected AFLP primer–enzyme combinations, and 825 (47.7%) of them were polymorphic (Table 1) ranging in size from 52 to 493 bp. Error rates were between 0.7% and 2.3% per primer–enzyme combination, with a mean of 1.2%. All individuals studied exhibited unique AFLP profiles. The final data set of the French samples consisted of 411 polymorphic loci in 299 individuals, whilst the data set of the Swiss samples consisted of 354 polymorphic loci in 322 individuals. We identified 254 (31%) common polymorphic loci between the two regions. Logistic regressions used to estimate independence between loci were significant for only 3.5% of all comparisons at \( P = 10^{-7} \), defined by using FDR.

**Population structure, historical and spatial effects**

*structure* estimated the maximal log-likelihood for all \( K \) values in the admixture model (Fig. 2a), but determining \( K \) only from the highest log probability of the data given \( K, Pr(X|K) \), was difficult. The \( \Delta K \) criterion (Evanno et al. 2005) reached its modal value for \( K = 2 \), suggesting that the uppermost level of genetic structure has two distinct clusters: the French and the Swiss Alps (Fig. 2b). The results for \( K = 2 \) to \( K = 4 \) (Fig. 2c) are presented to illustrate the formation of populations. No secondary contact zone between sampling regions was detected in our study (Fig. 2b). French and Swiss samples of *A. alpina* were significantly differentiated \( (F_{ST} = 0.1652, P < 0.0001) \). Based on these results, we considered the French and the Swiss Alps as two independent study regions, representing true replicates. The correlation between pairwise kinship coefficients and untransformed and log-transformed Euclidian distances between locations were significant for the French Alps \( (r^2 = 0.33, P < 0.0001 \).
and $r^2 = 0.39$, $P < 0.0001$, respectively) and for the Swiss Alps ($r^2 = 0.15$, $P < 0.0001$ and $r^2 = 0.16$, $P < 0.0001$ respectively; Fig. 3). The slope of the regression was steeper, indicating a stronger isolation by distance pattern in the French than in the Swiss Alps.

Detection of ecologically relevant loci using GEE

GEE were used to separately correlate allele frequencies in each of the two independent regions, the French and the Swiss Alps, to environmental variables. Thresholds...
transformed geographical distances. Points: (a) untransformed geographical distances and (b) log-transformed geographical distances.

**Fig. 3** Correlation between pairwise kinship ($F_{ij}$) coefficients of *Arabis alpina* individuals from 99 French (F, black line, black points) and 109 Swiss sampling locations (CH, grey line, grey points): (a) untransformed geographical distances and (b) log-transformed geographical distances.

for significant $P$ values were estimated at 0.00042 for the French and 0.00012 for the Swiss samples. For each locus, the best GEE model with at least one significant regression coefficient is given in Table S2 (Supporting Information). In the French data set, 61 loci (15% of all loci) were significantly correlated with one, two or three environmental variables (Table S2, Supporting Information), with 10.5% of these 61 loci showing non-independence, i.e. linkage disequilibrium (results from logistic regressions, see above). Among these 61 ERLs, $t_{\text{min}}$ was correlated with 21 loci (28%), $t_{\text{wi}}$ with 12 loci (14%), and $s_{\text{lp}}$ with six loci (7%; Fig. 4). In the Swiss data set, 21 loci (6% of all loci) were significantly correlated with one or two environmental variables (Table S2, Supporting Information), with 14.3% of these 21 loci showing non-independence. Of the 21 ERLs identified in this second region, $t_{\text{min}}$ was correlated with 10 loci (40%), $p_{\text{prcp}}$ with eight loci (36%), $t_{\text{wi}}$ with three loci and $s_{\text{lp}}$ with two loci (12%; Fig. 4). Second-order polynomials of environmental variables were involved in 32 loci in the French data set and in 12 loci in the Swiss data set (Fig. 4).

Four loci were jointly identified as ERLs in both regions studied (Table 2). Locus PM251.7 was correlated with $t_{\text{min}}$ both in the French and the Swiss Alps. The frequency of this locus increased with $t_{\text{min}}$ in the same order of magnitude in both regions as indicated by the positive value of the regression coefficients ($\beta_1 = 0.0374$ and 0.0305, respectively; Table 2) and by the spatial distribution of $t_{\text{min}}$ in the study area.

**Table 2** Generalized estimating equation (GEE) analyses for four amplified fragment length polymorphism (AFLP) loci in *Arabis alpina* responding to environmental variables in both the French (F) and the Swiss (CH) Alps. QIC is the quasi-likelihood under the independence model criterion for GEE. These best models were of the kind $\beta_0 + \beta_1 \text{Variable}_1$.

<table>
<thead>
<tr>
<th>Locus</th>
<th>Region</th>
<th>QIC</th>
<th>$\beta_0$</th>
<th>$\beta_1$</th>
<th>$P$-value</th>
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<tr>
<td>PM251.7</td>
<td>F</td>
<td>95.1</td>
<td>-13.7</td>
<td>0.0374</td>
<td>$8.98 \times 10^{-6}$</td>
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<td></td>
<td>CH</td>
<td>256.3</td>
<td>-6.2</td>
<td>0.0305</td>
<td>$1.33 \times 10^{-3}$</td>
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<tr>
<td>PM342.7</td>
<td>F</td>
<td>121.4</td>
<td>-15.4</td>
<td>2.6050</td>
<td>$4.18 \times 10^{-3}$</td>
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<tr>
<td></td>
<td>CH</td>
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<td>$1.08 \times 10^{-3}$</td>
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<tr>
<td>PM212.8</td>
<td>F</td>
<td>92.0</td>
<td>-25.2</td>
<td>2.8016</td>
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<td>EM125.1</td>
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<td>7.62 $\times 10^{-7}$</td>
<td>$9.72 \times 10^{-3}$</td>
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(Fig. 5a). For this locus, tmin was significantly smaller when the restriction fragment was present than when it was absent (t test, \( t = 6.7279, \text{df} = 119, P = 1.899 \times 10^{-6} \) in the French Alps and \( t = 8.6567, \text{df} = 124, P = 2.158 \times 10^{-14} \) in the Swiss Alps; Fig. 5b). PM342.7 and PM212.8 were both correlated with twi and prcp. The fourth locus, EM125.1, was correlated with twi and tmin. Pairwise correlations between the distributions of these four ERLs showed that the distribution of locus PM251.7 was significantly correlated to the distribution of locus EM125.1 (logistic regression, \( P = 3 \times 10^{-6} \)). This association also holds true for the two loci PM342.7 and PM212.8 (logistic regression, \( P = 2 \times 10^{-7} \)).

**Discussion**

Divergent selection in relation to environmental variation is a key process in the evolution of populations. As a result of recent technological advances in molecular methods, studying local adaptation using genome scans has received considerable attention, which also helps to better understand the responses of organisms to global change.

In this study, we identified 78 ERLs whose distribution was significantly related to the environmental variables tested, in particular to temperature (43 loci in the French Alps and 10 in the Swiss Alps) and precipitation (21 loci in the French Alps and eight in the Swiss Alps). Four of these loci were identified in both regions considered as independent spatial replicates. Clinal patterns of genetic variation observed in loci along environmental gradients have usually been interpreted as being caused by natural selection (Endler 1977; Schmidt et al. 2008). Therefore, such loci have been interpreted as being located in or being linked to candidate genes. We have strong confidence in the four loci...
identified in our study, because they were found to be of ecological relevance in two independent regions. However, we lack the experimental evidence to securely interpret these loci as being genomic regions under selection, although the use of an appropriate statistical approach gives strong support to this conclusion. Here, we will first discuss the innovative character and the reliability of our approach, but also its limitations, before interpreting our ERLs in an ecological context.

Previous studies on ERLs in plant species found similar patterns of genetic variation in allozyme frequencies along environmental gradients (Allard et al. 1993; Linhart & Grant 1996; Prentice et al. 2000). For example, Hirao & Kudo (2008) found a correlation between allozyme frequency and a flowering time trait in Primula cuneifolia along a snowmelt gradient. Shimono et al. (2009) identified morphological traits covarying with allozyme frequencies of fellfield and snowbed populations in Potentilla matsumurae, suggesting that the timing of snowmelt applies selective pressure, thus driving local adaptation in this alpine plant species. Recent technological advances have made large-scale genome scans available, optimizing the identification of loci potentially of ecological relevance. This is particularly important for our study, in which we tested the excessively large number of 825, mostly unlinked AFLP loci for their relationship with environmental variation.

We analysed our data based on individuals (typically three individuals per site), since we aimed at including many locations and a broad range of environmental conditions. In consequence, we did not use recent population genetic approaches searching for correlations between allele frequencies and climate variables (Yu et al. 2006; Hancock et al. 2008; Manel et al. 2010b), which are able to correct for population structure. Instead, we applied a correlative individual-based approach, GEE, including the possibility to detect false positive signatures of selection due to historical or spatial effects. Therefore, we used additional biological criteria to separate true selection signatures from those of population structure or history. Firstly, the population structure analysis showed that the French and Swiss regions can be considered as independent from a historical perspective, i.e. genetic drift exceeds migration. Both regions were characterized by significant isolation by distance, with gene flow occurring over small distances and decreasing over larger geographic distances. Identifying joint ERLs in the two independent and well differentiated regions ($F_{ST} = 0.1652$) is a way of preventing interference from purely historical effects (Endler 1986; Vasmagi 2006; Schmidt et al. 2008); an ideal situation (genome-wide vs. locus-specific clines) for detecting adaptive genetic variation. Secondly, our study regions did not represent admixture zones according to STRUCTURE analysis (Fig. 2), although previous phylogeographic analyses of A. alpina showed that the Swiss sampling area was in the vicinity of a contact zone of two evolutionary lineages in the central Alps (Assefa et al. 2007; Ehrich et al. 2007; Ansell et al. 2008). Finally, the isolation by distance patterns detected in both regions (Fig. 3) suggested spatial autocorrelation between sampling locations, for which GEE does not correct. The consequence of spatial autocorrelation between sampling locations in regression models has recently been discussed by Diniz-Filho et al. (2009) in the context of genetic data: interdependence among samples disturbs the significance of tests and reduces the power of the analysis. In our case, however, spatial autocorrelation decreased rapidly and was not relevant for comparisons of distant locations, reflecting recurrent gene flow only between close locations.

The choice, scale and precision of the environmental variables to be included are crucial criteria in landscape genomic studies. In our case, only variables showing a substantial gradient, or at least large variation in space, were included in the analysis to detect adaptive genetic variation (Etterson 2004). As a result, we did not consider different substrate types, which have been shown to be an important driver of genetic patterns in alpine plants (Choler et al. 2004; Alvarez et al. 2009); in any case Arabis alpina mainly occurs on calcareous substrate. On small spatial scales, it is appropriate to use finer measurements of environmental variables (e.g. Linhart & Grant 1996), but it is not always possible to obtain local measurements over extended areas. Therefore, climatic and topographic variables obtained from GIS databases are relevant alternatives (Joost et al. 2007; Manel et al. in revision).

Out of the 78 ERLs identified in the two study regions, four loci were correlated with at least one environmental variable in both the French and the Swiss Alps. One of them (locus FM251.7) was correlated with tmin in both study regions and showed higher allele occurrence when minimum temperatures were smaller (Fig. 5b). The spatial distribution of allele presence at this locus and of minimum temperatures in the French and the Swiss Alps showed that in France, the tmin effect was restricted to the Chartreuse, Cerces and Thabor areas (Fig. 5a), suggesting local adaptation on the massif scale, whereas the restriction fragment was present at the coldest sites, i.e. at the highest altitudes in Switzerland. The scale effect of adaptive genetic variation (i.e. local versus large scale) is discussed in a recent analysis on A. alpina (Manel et al. 2010a). The three other ERLs were not correlated with the same environmental variable in both study regions. As adaptation may include complex genetic responses, these markers might nevertheless show signs of selection. Linkage
disequilibrium between some pairs of the four loci suggested that loci PM251.7 and EM125.1 and loci PM342.7 and PM212.8 might be located on the same linkage groups, respectively, being associated with the same environmental variables except for EM125.1 linked to twi in the French Alps.

In our study, the environmental variable that explained most of the allele distributions was Tmin (Table S2, Supporting information). Minimal temperatures are a well-known selective pressure in mountain and alpine plants. Indeed, freezing temperatures constitute a major limitation to plant growth (Körner 2003), productivity and distribution (Boyer 1982). Freezing tolerance is a complex process involving a number of biochemical and physiological adaptations under genetic control (Hughes & Dunn 1996). Precipitation was also often correlated with allele occurrence in our study, but a direct link of the identified markers of ecological relevance to drought is not obvious, as adaptation to drought involves hundreds of genes (Chaves et al. 2003) and as moisture availability is considered to be a complex environmental trait along altitudinal gradients (Körner 2007).

The next step would be to link the identified ERLs to traits potentially involved in adaptation to the corresponding selective pressure. For alpine plants, the physiological traits involved in adaptation should mainly be linked to key growth-related traits (Chapin et al. 1993), flowering time (Hirao & Kudo 2004) and vernalization (Hopkins et al. 2008). Genes under natural selection are studied in the related model plant Arabidopsis thaliana (Mitchell-Olds & Schmitt 2006; Chiang et al. 2009; Reinninga et al. 2009) and have started to be investigated in A. alpina (Amasino 2009; Wang et al. 2009). Some functional traits have been studied in a landscape perspective, especially with regard to the genetic control of flowering and the perennial life history trait (Johanson et al. 2000; Wang et al. 2009). Perpetual flowering (PEPI) genes contribute to three perennial traits because they limit the duration of flowering, facilitate a return to vegetative growth, prevent branches from undergoing transition to flower production, allow polycarpic growth and confer an adaptive response to low winter temperatures, which restricts flowering to spring. Such genes might well be involved in the adaptive response of A. alpina, as we found ERLs correlated to minimum temperatures, which in turn relate to the length of the flowering period in alpine environments (Körner 2007). Further studies should therefore attempt to identify the genomic regions of A. alpina containing the ERLs by sequencing the corresponding AFLP fragments and comparing these sequences with the A. thaliana genome (The Arabidopsis Genome Initiative 2000). This could give a first indication of the genes involved as well as their functionality (Nielsen 2005; Stinchcombe & Hoekstra 2008). Validation is necessary to determine if the identified ERLs are truly under divergent selection or only false positives, and we are currently developing experiments to investigate the influence of minimal temperatures in the adaptation of Arabis alpina.

In conclusion, our results demonstrate the potential of using genome-wide marker surveys in conjunction with allele distribution models to reveal signatures of natural selection along environmental gradients. However, combining the information from genome scans with other approaches (fitness variation along gradients, genomic characterization of ERLs, transplant experiments; Feder & Mitchell-Olds 2003; Kawecki & Ebert 2004; Holderegger et al. 2008) may substantiate the genetic and functional architecture behind complex adaptive traits.

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Supporting Information

Additional supporting information may be found in the online version of this article.

Table S1 Description of Arabis alpina sampling locations in the French and the Swiss Alps

Table S2 Results of GEE analyses in Arabis alpina from the French and the Swiss Alps

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