Unravelling the architecture of functional variability in wild populations of *Polygonum viviparum* L.

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**Summary**

1. Functional variability (FV) of populations can be decomposed into three main features: the individual variability of multiple traits, the strength of correlations between those traits and the main direction of these correlations, the latter two being known as ‘phenotypic integration’. Evolutionary biology has long recognized that FV in natural populations is key to determining potential evolutionary responses, but this topic has been little studied in functional ecology.

2. Here, we focus on the arctico-alpine perennial plant species *Polygonum viviparum* L. We used a comprehensive sampling of seven functional traits in 29 wild populations covering the whole environmental niche of the species. The niche of the species was captured by a temperature gradient, which separated alpine stressful habitats from species-rich, competitive subalpine ones. We sought to assess the relative roles of abiotic stress and biotic interactions in shaping different aspects of functional variation within and among populations, that is, the multi-trait variability, the strength of correlations between traits and the main directions of functional trade-offs.

3. Populations with the highest extent of functional variability were found in the warm end of the gradient, whereas populations exhibiting the strongest degree of phenotypic integration were located in sites with intermediate temperatures. This could reveal both the importance of environmental filtering and population demography in structuring FV. Interestingly, we found that the main axes of multivariate functional variation were radically different within and across population.

4. Although the proximate causes of FV structure remain uncertain, our study presents a robust methodology for the quantitative study of functional variability in connection with species’ niches. It also opens up new perspectives for the conceptual merging of intraspecific functional patterns with community ecology.

**Key-words:** alpine plants, ecological niche, functional traits, intraspecific variation, lines of least resistance, phenotypic integration, variance-covariance matrix

**Introduction**

Intraspecific phenotypic variability has recently emerged as an important topic in the field of plant community ecology (Violle et al. 2012). Several studies have shown that, contrary to previous expectations, plant functional traits that vary between species across environmental gradients and are related to community assembly could also be highly variable within species and even within populations (Shipley & Almeida-Cortez 2003; Albert *et al.* 2010b). Accounting for this variability has proven to be crucial in answering various questions in plant ecology (see Jung *et al.* 2010 for community assembly; de Bello *et al.* 2011 for diversity measures; De Frenne *et al.* 2011 for functional strategies). To date, the study of intraspecific phenotypic variability in community ecology has remained mainly univariate (i.e. traits were studied separately, Violle et al. 2012 but see Reich *et al.* 2003; Albert *et al.* 2010a), although it is the entire trait syndrome that influences individual’s fitness and can be linked with species’...
environmental niches (Reich et al. 2003; Wilson & Nussey 2010). This lack of knowledge of the multivariate structure of functional traits at intraspecific level is particularly embarrassing. Indeed, there has been wide recognition in the field of evolutionary quantitative genetics that the variability of single traits as well as the correlations between them at the population level can be key in driving local adaptation, shaping the boundaries of species’ niches and determining their evolutionary potential (Kirkpatrick & Barton 1997; Gomulkiewicz & Houle 2009; Lavergne et al. 2010).

In this paper, we use the term ‘functional variability’ (hereafter FV) to jointly refer to the amount of variance in multiple functionally related traits (i.e. single-trait variances) and to the pattern of covariation between these traits, this later characteristic being known as ‘phenotypic integration’ (Pigliucci 2003). The functional variability of a population can be summarized by its phenotypic variance-covariance matrix and visualized as an ellipsoid in a multidimensional trait space (Fig. 1). This ellipsoid has three main features: (i) the extent of functional variability (hereafter FV extent), which represents the overall amount of phenotypic variability, is the volume of the ellipsoid; (ii) the shape of functional variability (hereafter FV shape), measured as whether the ellipsoid is closer to a sphere or to a segment, which describes the strength of the correlations between the different traits (i.e. the intensity of phenotypic integration) and (iii) the direction of functional variability (hereafter FV direction), which represents the main direction of variation in the multi-trait phenotypic space, is the main direction of the ellipsoid.

Based on this methodology, studying the link between multi-trait intraspecific FV and the ecological niche can be broken down into three main questions.

First, concerning FV extent, it is crucial to understand how it varies within the niche from its core to its edge. Several hypotheses exist regarding the mechanisms driving FV extent. On the one hand, stressful abiotic environments should reduce intraspecific FV due to strong directional selective pressures resulting in the environmental filtering of adapted phenotypes (Keddy 1992; see Arnold et al. 2008 for the effect of selection on genetic variability). This kind of strong environmental filtering is frequently observed at the interspecific level in extremely arid or cold environments where functional diversity at the community level is reduced (Cowling et al. 1994; de Bello, Leps & Sebastia 2006). On the other hand, strong biotic interactions in species-rich communities could result in larger intraspecific FV. Indeed, in such diverse communities, many different kinds of competitors with varying ecological strategies and different functional traits are likely to be encountered by different individuals of a same species. This should drive divergent selection and character displacement in different directions for different individuals in order to reduce competition for resources with individuals from other species (Brown & Wilson 1956), thus resulting in a wider spectrum of functional strategies in the local population (Reich et al. 2003). In any case, the effect of biotic interactions should be more important for traits that are related to coexistence mechanisms than for traits involved in the tolerance to abiotic conditions.

Secondly, it is important to understand what are the drivers of FV shape, and in particular in which part of the environmental niche the most integrated phenotypes are found. Theory predicts that correlational selection should be the main driver of strongly integrated phenotypes (Arnold et al. 2008), even though other genetic mechanisms could also increase trait correlations (Armbuster & Schwaegerle 1996). At the intraspecific level, phenotypic integration in plants has mainly been studied on floral morphology, with the recognition that strong selective pressures imposed by pollinators are responsible for the high degree of integration in floral structures (Berg 1960; Ordano et al. 2008; Armbuster et al. 2009). Concerning vegetative traits, it has been observed that plant species living in harsh environments often exhibit suites of functional traits that are strongly correlated (Chapin, Autumn & Pugnaire 1993). Several experimental studies have supported this view at the intraspecific level. For example, Gianoli (2004) showed that traits related to resource acquisition and growth in Convolvulus arvensis are more tightly correlated when environmental stress increases, which might be due to stronger energetic trade-offs between...
several physiological functions (see also Schlichting 1989). According to these observations, we would expect that the most integrated populations be found at the niche edges, and particularly where abiotic conditions are limiting. However, Tonsor & Scheiner (2007) have found an opposite result in Arabidopsis thaliana, where the overall degree of phenotypic integration does not change with CO₂ availability.

Thirdly, examining FV direction provides interesting insights into the main drivers of functional trade-offs and the main axes of multivariate phenotypic variation at the population level. On the one hand, environmental factors could impose certain energetic constraints and thus settle trade-offs between several traits, resulting in natural selection shaping the main direction of phenotypic variation (Schluter 1996; Webb et al. 2010). This has been exemplified at interspecific level by the leaf economics spectrum, a single axis of variation that captures most of the variance in key foliar traits over thousands of plants from all around the world (Wright et al. 2004). However, if selection is the main driver of FV direction, there are no reasons why two populations that face different environments could not have different main axes of phenotypic variation. On the other hand, genetic factors like pleiotropic effects, random drift, asymmetric gene flow between source and sink populations or linkage disequilibrium between traits can increase correlations between certain pairs of traits and thus set the main directions of FV (Armbruster & Schwaegerle 1996; Gomulkiewicz & Houle 2009). In the case of extremely strong genetic control on FV direction, these directions should be the same among populations and within different populations (Sokal 1978; Armbruster & Schwaegerle 1996).

In this paper, we examine how these three different aspects of FV vary across the environmental niche of the widely distributed arctico-alpine plant species Polygonum viviparum L. Using robust statistical techniques borrowed from quantitative genetics, we studied the multivariate functional variability of different populations in natural conditions along an environmental gradient typical of alpine landscapes (i.e., temperature). We specifically ask the following questions:

1. How does the extent of intraspecific FV vary across spatial scales, that is, what is the importance of intra-population trait variability compared to interpopulation trait variability?
2. Does the extent of intraspecific FV increase from the warm to the cold edge of the species’ niche due to the shift from environments dominated by competition to environments dominated by environmental filtering?
3. Is phenotypic integration higher at the edges of the niche due to more stressful conditions that impose stronger energetic trade-offs?
4. Do different populations share the same FV direction? And how does these directions relate to the environmental gradients and to the main direction of FV at the interpopulation level?

Material and methods

StudY SPECIES AND SITE

We chose Polygonum viviparum L. as a model species because of its large environmental niche. This herbaceous perennial occurs in all arctico-alpine regions of the northern Hemisphere. In the Alps, it can be found from the montane belt (starting around 1000 m of altitude), where plant biomass is high and competition for light and nutrients severe, to the upper alpine level (ending at c.a. 3000 m a.s.l.), where the environment is harsher and physiological limitations are stronger (Körner 1999, see Appendix S2). It has a preference for relatively moist habitats. The species has the specificity of bearing both flowers and bulbils (clonal reproductive organs) on the flowering spike.

We studied the species in the central French Alps Guisane Valley (Fig. S1, Supporting information) where it occurs in a variety of ecological contexts (from forests dominated by Larix decidua Mill. to alpine screes). To maximize the environmental differences between sites (Albert et al. 2010c), we stratified the sampling design following two independent gradients known to have high impact on the physiology of alpine plants (Körner 1999): mean annual temperature and solar radiation in June. These two variables were selected from a set of climatic variables interpolated at 50-m resolution Aurelhy model, (Benichou & Le Breton 1987) extracted from all known occurrence points for P. viviparum in the Guisane valley (data collected by the National Botanical Alpine Conservatory, http://www.cbn-alpin.fr/). The selection was made by choosing two orthogonal gradients that strongly correlated with the two first axes of the principal component analysis conducted on this set of variables (results not shown).

Temperature was the main environmental gradient and the primary determinant for P. viviparum’s environmental niche in our study area (Thuiller et al. 2010; Boulangeat, Gravel & Thuiller 2012). This climatic variable acts on plant physiology and phenology, with colder sites being subject to more frequent frost events even during the summer and experiencing a shorter growing season. Temperature also plays an indirect biotic role in conjunction with soil by discriminating between warm productive species-rich habitats and cold unproductive species-poor habitats (Körner 1999). Using botanical surveys to estimate species richness per site as well as a spectral measure of overall biomass per area (NDVI, see Appendix S2), we confirmed that mean annual temperature was indeed positively correlated to both species richness ($R^2 = 0.10$, $P = 6e-5$) and biomass per area ($R^2 = 0.28$, $P = 0.0005$).

This led us to interpret the temperature gradient as a climatic gradient influencing plant physiology and phenology but also as a gradient discriminating between sites mainly dominated by biotic vs. abiotic constraints. Such a contrast between the limiting role of abiotic stress at the cold end of the distribution and the primary importance of biotic interactions at the warm end of the distribution has recently been confirmed for several alpine plant species, including P. viviparum (Boulangeat, Gravel & Thuiller 2012). In contrast, even if it is usually an important gradient for alpine vegetation and although we explicitly sampled along it, solar radiation did not explain any FV characteristic at the population level: its influence is therefore not discussed in the following of this article.

FIELD TRAIT MEASUREMENTS

We sampled 29 populations at altitudes ranging from 1500 m to 2950 m, covering a large proportion of the climatic space occupied by the species in the study area [99% of the temperature gradient and 53% of the radiation gradient, Fig. S1 (Supporting information)]. Measurements were made at each population’s
flowering peak in order to sample each population at the same phenological stage (July 2010). In each population (10×10 m), three subpopulations of 1×1 m were selected to represent contrasted microenvironmental conditions, using the same methodology as Albert et al.’s (2010b). In each subpopulation, the following functional traits were measured on five randomly selected individuals: maximum vegetative height (H_{max}, top of plant photosynthetic tissue); total length of the inflorescence (H_{rhio}); ratio of sexual reproduction (SEX), ratio of the length of flowers divided by the total length of the spike; leaf dry matter content (LDMC, the ratio of leaf dry mass over fresh mass); specific leaf area (SLA, the ratio of leaf surface over fresh mass); leaf nitrogen content (LNC, the percentage of nitrogen in the dry mass of the leaf) and carbon-to-nitrogen ratio (C:N, the ratio of carbon over nitrogen in the leaf dry mass). These traits relate to various aspects of plant functional strategy (Westoby, Falster & Moles 2002), like resource acquisition and growth rate (LDMC, SLA, LNC, C:N), ability for light competition (H_{max}) and reproductive effort (H_{rhio} and SEX). Foliar traits are known to be physiologically correlated due to leaf economics constraints (Wright et al. 2004) and are thus suited to studying phenotypic integration. However, energetic trade-offs could also arise at the whole plant level due to resource allocation conflicts between growth, longevity and reproduction (Chapin, Autumn & Pugnaire 1993; Enquist et al. 1999; see Diggle 1997 for allocation of P. viviparum); our decision to include H_{max}, H_{rhio} and SEX was intended to include this higher-level trade-off.

CHARACTERIZING FUNCTIONAL VARIABILITY IN WILD POPULATIONS

Overall trait variability

To quantify the extent of intraspecific functional variability in the whole data set and understand the structure of intraspecific FV across spatial scales, we first broke down the variability of each trait at different hierarchical levels using mixed effects regression models. To do this, we used intercept models with random effects corresponding to the different levels of hierarchy (i.e. population and subpopulation nested in population). We then extracted the percentage of variance explained by each hierarchical level for each trait. Variance components were estimated using restricted maximum likelihood (REML).

To get a finer understanding of trait variation across our study area, we also examined the response of all traits against the temperature gradient, using linear or quadratic models with the same random effects as mentioned previously to account for the hierarchical structure of the data set. P-values for such models were obtained by likelihood ratio tests, using an R function provided by Christopher Moore (http://blog.lib.umn.edu/moor0554/canoe-moore/2010/09/lmer_p_valueslrt.html).

To quantify FV extent for each population, all traits were transformed to have a mean of zero and a standard deviation of one across the whole sample. Thus, all traits have equal importance in the subsequent analyses. For each population, a variance-covariance matrix for the seven traits was built (P-matrices). Overall trait variability (i.e. FV extent) in a population was measured as the trace of P (i.e. the sum of its diagonal elements), a measure commonly used for genetic variance matrices (Revell 2007).

Phenotypic integration: patterns and causes

A matrix of correlations between the seven traits was built for each population (P′-matrices), and the variance of the eigenvalues of P′ was taken as an index of integration (i.e. FV shape, Cheverud, Wagner & Dow 1989), higher values meaning stronger correlations between traits.

The direction of phenotypic integration was compared between populations by determining the axis of maximum phenotypic variation, P_{max} (first eigenvector of P, also known as ‘the line of least resistance’, Schluter 1996), for each population. For each couple of populations, one minus the correlation between their P_{max} was used to measure the functional distance between them, producing a matrix of functional distances between populations (P-dist).

To test whether and how environmental or genetic constraints drive FV direction, we compared P-dist to different environmental distance matrices (Euclidean distance on the climatic plane defined by temperature and radiation and Euclidean distance on the temperature gradient only) and geographical distances using Mantel tests. As the influence of gene flow between populations was expected to mainly play a role at small geographical scales (of the same order of magnitude as the species’ dispersal distance), we also used Mantel correlograms to unravel these small-scale dependencies.

We conducted the same analysis on FV direction using Random Skewers (Cheverud 1996) to measure functional distances between populations. Although Random Skewers were originally designed to compare the responses of different populations to putative selection events, they can also be used to compare all kinds of variance-covariance matrices (e.g. Kolbe et al. 2011) and have advantages over P_{max} methods in that they compare the properties of entire matrices. This additional procedure was used to back up the results obtained with the P_{max} analysis and led to the same conclusions (detailed method, R code and results for Random Skewers are available in Appendix S3).

ROBUSTNESS OF MATRIX ESTIMATION

Given our sampling implied a low number of measured individuals within each population (i.e. 15), it could impede robust estimation of the P and P′ of each population. We measured the robustness of matrix estimation using a bootstrapping procedure (Cheverud, Wagner & Dow 1989) and found that on average there are 71% of chances that differences between two P matrices are not meaningful and 14.3% of chances for P′ matrices (detailed description in Appendix S5). This uncertainty is however counter-balanced by the two main strengths of our approach which are that (i) we studied FV within and among numerous (i.e. 29) populations of the same species, thereby rendering our analyses less sensitive to this matrix estimation error, and (ii) followed a stratified hierarchical sampling along in situ and continuous environmental gradients. This should provide a more comprehensive picture of trait variability and integration across the whole niche of the study species than what is generally done under controlled conditions on few discrete environmental conditions.

Results

EXTENT OF TRAIT VARIABILITY

Variance decomposition revealed two different cases. In the case of vegetative height (H_{max}), most of the variation (73%) was found between populations. Conversely, for all other traits included in our study, around half of the variance occurred between individuals of the same subpopulation (Table 1). Overall, there was little variation between different subpopulations (1–21% depending on the trait). The subsequent analyses carried out at population level were then justified, as FV was rather high within populations.

All of the traits we studied except SEX showed a significant relationship with mean annual temperature (see
Fig. 2). Mean population values of LDMC and LNC decreased with temperature, while SLA, C:N and H\text{inflo} increased with temperature. H\text{max} showed a quadratic relationship, reaching a maximum value for intermediate temperatures.

FV extent (overall trait variability) in each population positively correlated with the mean annual temperature of the site ($R^2 = 18\%$, $P = 0.012$, see Fig. 3). No significant relationship was found with solar radiation.

**PHENOTYPIC INTEGRATION**

The strength of phenotypic integration was in general relatively high for all populations. Indeed, under the assumption of no correlation between the seven traits studied and given that we sampled 15 individuals per population, the expected value for the integration index is 0.4 (Wagner 1984). To evaluate a confidence interval for that value, we built a null distribution for the integration index by randomly sampling seven trait values for 15 individuals (Gaussian traits, 100 000 resamples) and computing the integration index. We obtained a 95% quantile of 0.75. Observed values of phenotypic integration across the 29 populations were always significantly stronger than randomly expected (min = 0.77) and were on average rather high (mean = 1.23).

This integration showed a triangular relationship with mean annual temperature (Fig. 3). Quantile regressions confirmed that the most integrated populations were found at the middle of the temperature gradient, which corresponds to the niche core: the 75% percentile of the distribution of integration values shows a quadratic relationship with temperature ($P$-value = 0.022).

The main directions of phenotypic integration, estimated by $P_{\text{max}}$, generally correlated between populations ($\text{cor} = 0.48 \pm 0.28$). No general line of least resistance emerged although most of the $P_{\text{max}}$ were directed towards high variance in LNC. Interestingly, the main direction of phenotypic variation for all 29 populations pooled together is orthogonal to this dominant intrapopulation direction (results from a PCA, see Fig. 4).

The differences in the main directions of phenotypic integration between populations were not explained by environmental nor geographical distance, as all the Mantel tests were non-significant ($P$-values = 0.328; 0.793 and 0.668 for environmental, temperature and geographical distances, respectively). However, FV direction in

**Table 1. Variance decomposition of the different traits. The percentage of variance explained by the different hierarchical levels is shown for each of the seven functional traits. Coefficients of variation are presented in the last row.**

<table>
<thead>
<tr>
<th>Hierarchical level/Trait</th>
<th>$H_{\text{max}}$</th>
<th>SEX</th>
<th>$H_{\text{inflo}}$</th>
<th>LDMC</th>
<th>SLA</th>
<th>C:N</th>
<th>LNC</th>
</tr>
</thead>
<tbody>
<tr>
<td>Population</td>
<td>73</td>
<td>17</td>
<td>46</td>
<td>42</td>
<td>56</td>
<td>38</td>
<td>42</td>
</tr>
<tr>
<td>Subpopulation</td>
<td>6</td>
<td>21</td>
<td>8</td>
<td>6</td>
<td>10</td>
<td>1</td>
<td>3</td>
</tr>
<tr>
<td>Individuals</td>
<td>21</td>
<td>62</td>
<td>46</td>
<td>51</td>
<td>34</td>
<td>61</td>
<td>55</td>
</tr>
<tr>
<td>Coefficient of variation</td>
<td>0.43</td>
<td>1.52</td>
<td>0.34</td>
<td>0.14</td>
<td>0.28</td>
<td>0.26</td>
<td>0.24</td>
</tr>
</tbody>
</table>

Fig. 2. Response of functional traits to the temperature gradient. Individual trait measures for all traits except SEX are plotted in grey. Black lines show the regression lines (quadratic regression in the case of $H_{\text{max}}$).
populations tended to be positively correlated at short distances (<200 m, cor = 0.83, P = 0.043, Fig. S4 (Supporting information)), but no relation between functional and geographical distances was found for larger distances.

Discussion

Our study reveals some general patterns of functional variability in *P. viviparum*. The primary observation is that trait values are highly variable and that a large proportion of this variation is found between individuals of the same population, confirming previous observations on plants (Albert et al. 2010b; Messier, McGill & Lechowicz 2010). This could be due to high levels of phenotypic plasticity, but the fact that our attempt to distinguish subpopulations does not explain much FV may also reveal that microenvironmental heterogeneity plays a role at a smaller scale than the one we chose (11 m), possibly at individual scale. This highlights the importance of studying FV within populations. However, populations do exhibit some differences, and temperature strongly influences mean trait values at the population level, as already observed for various types of alpine plants including *P. viviparum* (Albert et al. 2010b). High values of LDMC and LNC along with
low values of $H_{\text{max}}$ and SLA for populations at the cold end of the gradient are characteristic of stressful environments (Chapin, Autumn & Pugnaire 1993) and indicate cold tolerance in these populations (Reich et al. 2003).

When trying to understand how FV extent is structured across *P. viviparum*’s niche, we found that the overall trait variability increases with mean temperature. This pattern supports the hypothesis that habitat filtering prevails in environments dominated by abiotic constraints (cold edge of the niche) and that functional divergence may be high within communities with a high number of interspecific biotic interactions (warm edge of the niche). Such a pattern of increased trait variability in richer communities has already been observed for morphological traits in grasshoppers of the genus *Melanoplus* (Roff & Mousseau 2005). In the case of *P. viviparum*, the very harsh conditions experienced in alpine habitats should lead to strong directional selection pressures for increased resistance to cold, drought and high solar radiation, whereas in the warmer subalpine meadows plant competition is expected to be stronger and lead to phenotypic divergence for niche partitioning between interacting individuals, thus resulting in increased trait variance within species (Weiher & Keddy 1995; Cornwell, Schwilk & Ackerly 2006; but see Spasojevic & Suding 2011).

We also observed that phenotypic integration is in general relatively high within *P. viviparum* populations. This is primarily due to the strong correlations between the four foliar traits included in this study (cor $= 0.45 \pm 0.32$ in absolute value over the 29 populations), which are known to reflect the world-wide leaf economics spectrum (Wright et al. 2004). The allometric correlation between $H_{\text{max}}$ and $H_{\text{info}}$ explains the rest of this pattern. Yet, no systematic trade-off was detected between reproductive, growth and persistence functions (the mean correlation for $H_{\text{info}}$ with any of the foliar traits is always $< 0.08$ in absolute value).

Note that the allocation to sexual reproduction (SEX) is a very idiosyncratic trait that does not correlate to the environmental gradients in our study area (contrary to what has been observed in the Arctic by Dormann, Albon & Woodin 2002), nor to any other trait measured.

One important finding is that the most integrated populations are found at the centre of the niche (i.e. middle of the temperature gradient, see Fig. 3). This result is contrary to our expectations and contrasts with some studies in controlled conditions where phenotypic integration has been found to increase with stress (e.g. Schlichting 1989; Gianoli 2004). We suggest that this pattern could be due to demographic asymmetries between the centre and the margins of the niche. Indeed, many theoretical models suggest that the larger population sizes at the centre of the niche lead to a better response to natural selection (e.g. Kirkpatrick & Barton 1997), which could produce integrated, more ‘optimized’ phenotypes. Conversely, marginal populations could be subject to high levels of both genetic drift and gene flow from the central populations, rendering selection inefficient and leading to low integration (Sexton et al. 2009). In our study area, we verified that population size is on average higher in the middle of the temperature gradient (F. Boucher, field observation).

The last attribute of FV that we intended to study was its direction. We found that the main axis of trait variation within populations is often related to variance in leaf nitrogen content, a trait linked to soil nitrogen uptake efficiency in fertile environments (Zatylny & St-Pierre 2006) and which also strongly affects the plant’s photosynthetic rate (Reich, Walters & Ellsworth 1991). This high variation in LNC within populations might be explained by the heterogeneity of the nitrogen supply in soils. This heterogeneity is both qualitative and quantitative: nitrogen can be present either in its organic form, which is costly to acquire, or in the form of ammonia or nitrates, and the amount of...
these alternative forms varies spatially. Indeed, it has been revealed that fine-scale factors like soil characteristics have a great influence on the functional diversity of plant communities in the Guisane valley (Bello et al. in press). The slightly lower variances in LNC and C:N found in colder populations of *P. viviparum* could be due to the predominance of organic nitrogen in high altitude sites (Averill & Finzi 2011) or to less spatial heterogeneity in soil nitrogen concentrations in these habitats, relative to subalpine ones. Interestingly, the main direction of phenotypic variation at the interpopulation level is almost orthogonal to this general intrapopulation direction and lines up with traits that are more directly related to the abiotic environment, supporting the main axis of variation observed at the intraspecific level in plants. Indeed, $H_{\text{max}}$ strongly correlates with temperature in our study area (Fig. 2), ranging from more than 50 cm in subalpine meadows to less than 5 cm in the alpine sites. LDMC is also strongly affected by climatic changes because it is expected to increase leaf longevity and thus resource conservation, which are likely to be favoured in stressful habitats. This suggests that the main functional trade-offs revealed for plants at the intraspecific level (e.g. leaf economics spectrum) over large geographical gradients might not be reflected at finer spatial scales (e.g. population level). This finding could have profound implications for the study of local coexistence in community ecology. A similar result has already been found for two forest herb species that show an opposite pattern of correlation between plant height and seed mass at the intraspecific level than the one observed at the intraspecific level (De Frenne et al. 2011). The fact that environmental distances do not correlate with functional distances between populations (measured either by $P_{\text{max}}$ or by Random Skewers correlations) confirms that climate is not the main driver of FV direction at the population level. On the contrary, the strong spatial autocorrelation that we found at small distances in the functional structure of populations suggests that high genetic similarity between close populations could result in very similar integration patterns (Stone, Nee & Felsenstein 2011). However, the fact that the main direction of integration differs significantly between populations shows that genetic correlations between traits are not excessively strong (Armbruster & Schwaegerle 1996).

Given that additive genetic variances were not measured and that the level of heritability for each studied trait is uncertain (Ackerly et al. 2000; Geber & Griffen 2003), any potential evolutionary interpretations of the patterns we report must be cautious. It is indeed possible that differences in intraspecific FV structure are only due to differences in genetic diversity in our populations arising from past demographic fluctuations (Wright 1969). However, these differences might as well be largely due to phenotypic plasticity (Pigliucci 2003) and in particular to the fact that individuals living in high-resource environments are expected to be more plastic than their stressed conspecifics (Chapin, Autumn & Pugnaire 1993; Grassein, Till-Bottraud & Lavorel 2010). Integrating direct measurements of genetic diversity and the relatedness of populations in the kind of ecological study proposed herein might constitute a promising avenue for future research (e.g. Martin, Chauvais & Goudet 2008) and will help to disentangle the relative effects of ecology, demography and genetics on the functional variability of populations.

**Conclusion**

Put together, our results give a clearer picture of how intraspecific FV is structured in different parts of *P. viviparum*’s environmental niche (Fig. 5). Indeed, populations at the ‘cold end’ of the niche have low FV but are mildly integrated. Populations located at the centre of the niche have intermediate trait variability and varying degrees of integration, but some are subject to strong trade-offs between traits. Finally, populations of the ‘warm end’ of the niche are both highly variable and little integrated. Our study thus shows the importance of the environment in setting FV extent and reveal substantial asymmetry in the species’ environmental niche between its ‘biotic’ and ‘abiotic’ edges. Interestingly enough, this finding corroborates hypotheses and observations traditionally made in community ecology that strong abiotic filters lead to greater functional convergence between species coexisting within natural communities (Webb et al. 2002). Finally, we demonstrate that the main functional trade-offs differ within and among populations and that the idea that phenotypic integration increases in stressful environments cannot be considered a rule.

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**References**


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Supporting Information
Additional Supporting Information may be found in the online version of this article:

Fig. S1. 3D map of the Guisane valley, stretching from Briançon (1200 m a.s.l.) to the Pic Blanc du Galibier (2955 m a.s.l.) in the inner French Alps. The Lautaret pass is located at 2050 m and the Guisane river is drawn in white. The 29 sampled populations of P. viviparum are shown in black dots. The map is colored according to the mean annual temperature (from white to black when going to colder sites, source: Aurelhy model [Benichou & Le Breton 1987]). The small plot in the bottom-left corner shows the position of the sampled populations (black dots) and of all populations of P. viviparum of the Guisane Valley recorded by botanists of the CBNA (grey dots) in the climatic plan defined by mean annual temperature and solar radiation in June. The temperature gradient is particularly well covered (99%).

Fig. S2. Plot of NDVI vs. mean annual temperature for 36 populations (including the 29 presented in this study) of P. viviparum in the Guisane Valley. The regression line is drawn in red.

Fig. S3. Relation between plant species richness and mean annual temperature in the Guisane valley.

Fig. S4. Correlograms for P-dist (left) and RS-dist (right), black squares meaning that the values of correlations are significantly different from zero and the first point corresponding to all pairs of populations that are closer than 200 m.

Fig. S5. Results of a PCA lead on all ‘lines of least resistance’: general metapopulation value (PV_total, circled in red), observed population values and bootstrapped values (2930 vectors). P_max are plotted on the two first axes of the PCA. The bottom-left panel shows how the seven traits relate to these axes.

Table S1. P-values of the Mantel tests.
Table S2. P-values of the partial mantel tests between functional distances and environmental distances, where geographic distance between populations was used as a covariate.