

**Intra-individual variation in *Galium odoratum* is partly genetically based and is affected by experimental drought and shading**

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Running title: Intra-individual variation under experimental drought and shading

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# 1 **Abstract**

## 2 *Background and aims*

3 Climate-change induced warmer spring temperatures advance tree leaf-out and result in  
4 earlier shading of the forest floor. Climate change also leads to more frequent droughts. Forest  
5 understorey herbs may respond to these environmental changes by varying functional traits at  
6 different hierarchical levels of organisation. While trait variation at the intra-specific level is  
7 well-studied, little is known about how variation at the intra-individual level responds to  
8 environmental changes.

## 9 *Methods*

10 We sampled genets of the forest understorey herb *Galium odoratum* from 21 populations in  
11 three regions in Germany, varying in microclimatic conditions. The genets were transplanted  
12 into a common garden, where we applied shading and drought treatments. We measured plant  
13 height, leaf length and width, and calculated the coefficient of variation (CV) at different  
14 hierarchical levels: intra-population, intra-genet, intra-clone and intra-ramet, the latter two  
15 representing intra-individual variation.

## 16 *Key results*

17 Variance partitioning showed that intra-ramet CV explained most of the total variation,  
18 followed by intra-clone CV. We found significant variation in CV of plant height and leaf  
19 width among populations of origin, indicating that CV is at least partly genetically based.  
20 Mean soil temperature at population origins correlated negatively with CV in plant height,  
21 suggesting adaptation to local conditions. Furthermore, we observed that early shade led to  
22 increased intra-clone CV in leaf length and drought reduced intra-ramet CV in leaf width.  
23 Finally, intra-ramet mean leaf width and CV were independent under control conditions but  
24 became correlated under drought.

1 *Conclusions*

2 Our experimental results reveal genetically based patterns in CV and correlations with soil  
3 temperature, indicating that intra-individual variation can evolve and may be adaptive. Intra-  
4 individual variation responded plastically to drought and shading, suggesting functional  
5 changes to improve light capture and reduce evapotranspiration. In conclusion, intra-  
6 individual variation makes up the majority of total trait variation and can play a key role in  
7 plant adaptation to climatic change.

8

9 **Keywords**

10 Clonal plant, common-garden experiment, forest understorey herbs, *Galium odoratum*,  
11 genetic differentiation, intra-individual variation, intra-specific trait variation, microclimate,  
12 phenotypic plasticity, sub-individual variation

## 1 **Introduction**

2 Intra-specific trait variation, *i.e.*, the phenotypic variation among individuals of the same  
3 species, can have important ecological impacts from individual functioning to population  
4 dynamics and species distributions, and from species interactions (Westerband *et al.*, 2021) to  
5 co-evolutionary dynamics and community assembly (Violle *et al.*, 2012; Westerband *et al.*,  
6 2021). Intra-specific trait variation is generated through genetic variation and phenotypic  
7 plasticity (Albert *et al.*, 2010; Des Roches *et al.*, 2018; Westerband *et al.*, 2021), and it has  
8 repeatedly been shown that intra-specific trait variation can be comparable in magnitude to  
9 inter-specific variation (Kinnison and Hairston, 2007; Siefert *et al.*, 2015; Govaert *et al.*,  
10 2016).

11 In clonal plant species, one genet can produce several ramets (*i.e.*, a potentially  
12 independent individual derived by vegetative production from a single parent plant) that can  
13 vary phenotypically, although arising from the same genotype (Sobral *et al.*, 2013). Such  
14 phenotypic variation among repeated homologous structures within the same individual is  
15 known as intra-individual variation (Herrera, 2009). In clonal plants, the intra-individual  
16 hierarchical level upwards from the ramet is the clone, followed by the intra-specific levels;  
17 genet and population (Fig. 1). Increasing evidence shows that intra-individual variation may  
18 have effects on the overall individual fitness and underlying functional mechanisms and may  
19 affect population-level variation (Herrera *et al.*, 2015; Herrera, 2017; March-Salas *et al.*,  
20 2021). Due to the modular nature of plants, the majority of variation in homologous structures  
21 of plants (e.g., fruits, leaves and seeds) may in fact be explained at the intra-individual level  
22 (Herrera, 2017). For instance, Herrera *et al.* (2015) found that the intra-ramet level explained  
23 approximately 50% of both leaf length and leaf width, while the intra-specific level explained  
24 less than 40% for both traits, and the intra-clone level explained 5-10%. Therefore, if intra-

1 individual variation is not considered, large fractions of functional variation and their effects  
2 on fitness could potentially be overlooked.

3         The part of phenotypic variation that is genetically based can be the result of mutation,  
4 genetic drift, gene flow, or natural selection, with only the latter causing adaptive evolution  
5 (Hedrick, 2009). Clonal plants can experience natural selection on traits at several levels –  
6 genet, clone and ramet – rendering measurements of natural selection and predictions of  
7 evolutionary responses considerably complex (Fischer and van Kleunen, 2001). Moreover,  
8 previous studies have shown that intra-individual variation can be subject to selection within  
9 each of these levels (Sobral *et al.*, 2013; Alonso *et al.*, 2018; March-Salas *et al.*, 2021). This is  
10 important because a genetic basis for intra-individual variation would be the first step to prove  
11 that intra-individual variation can have a functional role. However, the vast majority of  
12 studies on intra-individual variation has been conducted under natural conditions, impeding  
13 the assessment of the genetic basis of intra-individual variation. A commonly used and  
14 powerful approach to test for genetic inheritance of traits is to grow plants in a common  
15 garden (Turesson, 1922; De Frenne *et al.*, 2013; Scheepens *et al.*, 2013). Although this  
16 approach commonly focusses on mean trait values at the population level and is increasingly  
17 being used to study genetic differentiation in the strength of phenotypic plasticity to  
18 experimental treatments (West-Eberhard, 2003; Forsman, 2015), the common garden  
19 approach can also be used to measure genetic variation in intra-individual variation and,  
20 likewise, in the plastic responses of intra-individual variation to any applied treatments.

21         While genetic adaptation is important for population survival on the long run, plastic  
22 responses are crucial for plant survival to rapid changes in environmental conditions,  
23 including those provoked by climate change, although they have their limits. In Concord,  
24 North America, tree species have significantly advanced their leaf-out with nearly two weeks  
25 since the 1850s in response to increasing spring temperatures, whereas forest understorey

1 herbs were much less responsive (Heberling *et al.*, 2019). This phenological mismatch  
2 between overstorey trees and understorey herbs is expected to increase with further warming  
3 (Chen *et al.*, 2019; Vitasse *et al.*, 2022), suggesting that plastic responses may not be  
4 sufficient over time and genetic adaptation in trait means and/or plasticity is needed.  
5 Additionally, droughts are becoming more frequent with global climate change (Dai, 2013;  
6 Büntgen *et al.*, 2021), also in temperate forests (Trenberth *et al.*, 2014; Millar and  
7 Stephenson, 2015; Dai *et al.*, 2018), adding yet another stressful condition for understorey  
8 plants to overcome. Plants have evolved different ways to cope with drought: they can either  
9 escape, avoid or tolerate drought stress (Fang and Xiong, 2015). In response to drought  
10 conditions, plants can accelerate their reproductive cycle (escape strategy) (Franks *et al.*,  
11 2007), reduce their mean leaf size and increase specific leaf area to lower their transpiration  
12 rate (avoidance strategy) (Esau, 1960) or improve water use efficiency (tolerance strategy)  
13 (Marron *et al.*, 2003; Liu and Stützel, 2004; Wellstein *et al.*, 2017).

14 In the above examples, population responses may not only harbour changes in trait  
15 means but also in intra-individual variation, for instance in leaf morphology. Larger variation  
16 in leaf characteristics and architecture may allow plants to thrive under variable light and  
17 drought conditions. Indeed, capturing different light intensities may help plants to acclimate  
18 to seasonal variation in light conditions, as occurs in the deciduous forest understorey  
19 (Martinez and Fridley, 2018). Thus, functional variation within the same individual, such as  
20 physiological and morphological variation among leaves, may help plants to adapt to fine-  
21 grained environmental changes and ultimately favour population performance (Winn, 1996a).  
22 Nevertheless, experiments addressing the function of intra-individual level are to our  
23 knowledge absent.

24 Here we examine whether and how inter-individual and intra-individual variation is  
25 determined by genetic variation, correlates with microclimatic conditions, and responds to

1 experimental environmental manipulations, *i.e.*, drought and shading treatments. We used the  
2 understory clonal herb *Galium odoratum* (L.) Scop. as study species to 1) compare variation  
3 within various hierarchical levels of organisation (population, genet, clone and ramet),  
4 measured in three functional vegetative traits: plant height, and leaf length and width; 2) to  
5 test whether intra-specific and intra-individual trait variation has a genetic basis; 3) to test if  
6 forest management and related micro-environmental factors at the site of origin may have  
7 affected genetic variation in individual trait variation; and lastly, 4) to investigate how  
8 experimentally induced drought and earlier shading drive plastic responses in individual trait  
9 variation and trait means. By answering these questions, this study provides insightful  
10 knowledge on the variation and evolutionary causes of intra-individual variation in the face of  
11 future climatic conditions.

12

## 13 **Material and Methods**

### 14 *Study species and experimental system*

15 *Galium odoratum* (L.) Scop. (Rubiaceae) is a perennial forest understorey herb, reaching a  
16 height of 10-30 centimetres. The leaves are lancet-shaped, widest in or just above the middle,  
17 and mostly appearing eight at a time arranged in whorls. *Galium odoratum* can reproduce  
18 sexually via seeds (outcrossing), but also heavily relies on vegetative spread through stolons  
19 (Frederiksen and Rasmussen, 2006).

20 We sampled plants from forest plots (100 m × 100 m) in the Biodiversity Exploratories  
21 ([www.biodiversity-exploratories.de](http://www.biodiversity-exploratories.de)), a large-scale platform for ecological research in  
22 Germany (Fischer *et al.*, 2010). In May 2020, we sampled individuals of *G. odoratum* in three  
23 different regions across Germany: Schwäbische Alb (ALB) characterized by calcareous  
24 bedrock with an elevation up to 860 m, an annual mean temperature of 6-7 °C, and a mean

1 annual precipitation of 700-1000 mm; Hainich-Dün (HAI) also consisting of calcareous  
2 bedrock, with a maximum elevation of 550 m, mean annual temperature of 6.5-8 °C, and a  
3 mean annual precipitation of 500-800 mm; and Schorfheide-Chorin (SCH) which is a young  
4 glacial landscape with a maximum elevation of 140 m, annual mean temperature of 8-8.5 °C,  
5 and mean annual precipitation of 500-600 mm (Fischer *et al.*, 2010). An index for silvicultural  
6 management intensity (SMI) has been developed to capture forest management intensity  
7 (Schall and Ammer, 2013, 2014). We chose nine plots along the SMI gradient from low to  
8 high management intensity in each region. In each plot, we sampled five individuals with a  
9 minimum inter-individual distance of 10 meters to ensure they were genetically different  
10 (hereafter referred to as genets). Each genet was separated into four ramets and planted into  
11 multitrays (51.5 cm width, 33.5 cm length, 5.5 cm deep, 54 cells per tray; potted with potting  
12 soil “CL T torffrei”, Einheitserde, Sinntal-Altengronau, Germany) for establishment and  
13 growth until November 2020, when all individuals were transferred into 1.5 L pots with  
14 potting soil (“Typ T, Struktur 1B”, Hawita, Vechta, Germany). In spring 2021, all pots were  
15 relocated to a foil tunnel allowing us to apply watering and shading treatments. Mortality in  
16 early spring caused reduced numbers before the start of the experiment (see Table S1 for final  
17 numbers).

18         The ability of *G. odoratum* to spread vegetatively allows us to study numerous  
19 hierarchical levels of organisation: Population/Genet/Clone/Ramet (Fig. 1). We define ramet  
20 as a single shoot, and thus a potentially independent plant unit (usually multiple in a pot),  
21 clone as all connected ramets (*i.e.*, in a single pot), and genet as all plants which are  
22 genetically identical (*i.e.*, individuals that were sampled from the populations and  
23 subsequently divided into four different clones, each planted in a single pot).

24 *Shading and drought treatments*



1 To simulate the shaded forest understorey environment, we applied shading cloth over the foil  
2 tunnel in two layers (45% shading for each layer resulting in 90% total shading). Each layer  
3 of shading cloth was applied one week apart. To simulate future tree leaf-out conditions, the  
4 first layer was applied on 12 Apr. 2021, approximately two weeks before the anticipated leaf  
5 out of beech and oak trees in the surrounding area. Control shading cloth was applied when  
6 the natural leaf out of surrounding trees were observed in the area (Frankfurt am Main,  
7 Germany) on 30 Apr. 2021, likewise applied in two stages, one week apart.

8 We applied the drought treatment at the flowering start of the first plant that flowered  
9 (7 May 2021). The drought treatment was applied as a single event, in which all watering was  
10 ceased until substantial wilting (50% of all pots) was observed. This drought treatment lasted  
11 for two weeks until watering of all pots was resumed. Control plants received water by  
12 irrigation from above *ad libitum* during the whole experiment. The shading and drought  
13 treatments were applied in a full-factorial design (Control, N = 30 pots; Early shading, N =  
14 39; Drought, N = 37, Early shading + Drought, N = 29).

#### 15 *Measured traits and their variation*

16 We measured plant height, leaf length and leaf width after all plants finished flowering (end  
17 of June). Plant height was measured to the nearest 0.5 cm for all ramets in each pot. Leaf  
18 length and leaf width were measured with callipers to the nearest 0.1 mm precision on one  
19 randomly chosen leaf per whorl on up to five randomly chosen ramets per pot.

20 The coefficient of variation (CV) – calculated as the standard deviation divided by the  
21 mean of a specific trait – was used to estimate variation in each trait at each hierarchical level  
22 of organisation: intra-population ( $CV_{IP}$ ), intra-genet ( $CV_{IG}$ ), intra-clone ( $CV_{IC}$ ) and intra-  
23 ramet ( $CV_{IR}$ ) (Fig. 1). The  $CV_{IR}$  was not calculated for plant height since one ramet has only a  
24 single height value and thus no CV can be calculated.

## 1 *Data analyses*

2 Firstly, to test which hierarchical level of organisation explains the most variation in our  
3 measured traits, we performed variance partitioning on the traits using the R package ‘cati’  
4 and the function ‘partvar’ (Taudiere and Violle, 2016). Two different variance partitioning  
5 models were performed. The variance partitioning for plant height only included the nested  
6 structure of intra-population, intra-genet, and intra-clone. The variance partitioning for leaf  
7 length and leaf width included all hierarchical levels.

8         Secondly, we ran linear mixed-effect models (LMM) for  $CV_{IG}$ ,  $CV_{IC}$ , and  $CV_{IR}$  of  
9 each trait with treatments as fixed factor for intra-clone and intra-ramet level. We used the  
10 hierarchical levels of organisation as nested random factors, where the region and population  
11 level reflect genetic differentiation among populations, and the genet level is reflecting  
12 genetic differences among individuals. For intra-genet level we had no fixed factors and  
13 defined the intercept as +1. To test for the genetic basis of intra-individual variation, we  
14 performed Akaike information criterion (AIC) model selection. Each level of the nested  
15 random factor was removed one at a time, allowing us to compare the AIC of the various  
16 models. If the difference in the goodness of fit between two models exceeds an absolute value  
17 of 2 (Vrieze, 2012), we interpret the models as being significantly different and the  
18 hierarchical level of organization in the model with the lowest AIC value having importance  
19 for explaining CV in the trait and, since the data originate from plants grown under controlled  
20 conditions, is at least partly genetically based.

21         Thirdly, we tested for treatment effects on CV using LMMs with microclimatic  
22 variables and the SMI from the origin of each population as fixed effects, including shade,  
23 drought, and the interaction between the two treatments on the intra-clone and intra-ramet  
24 level. Hierarchical levels of organisation were used as random nested structure. LMMs were  
25 run using the function “lmer” from the package “lme4” (Bates *et al.*, 2007) and model test

1 results were obtained by applying the function “Anova” from the package “car” (Fox *et al.*,  
2 2012). To investigate significant differences between treatments with more than two levels or  
3 significant interactions, a post-hoc Tukey test was applied using the function ‘lsmeans’ from  
4 the package ‘lsmeans’ (Lenth, 2016).

5 Fourthly, in the same mixed-effects models, we tested for relationships between CV  
6 and environmental variables at the sites of origin. To avoid multicollinearity of environmental  
7 variables in the mixed-effects models, we applied a principal component analysis (PCA) from  
8 the package ‘factoextra’ on the following microclimate variables: mean annual soil  
9 temperature, mean annual air temperature, mean annual humidity, and mean annual  
10 precipitation. The PCA showed that PC1 explained 63.8% of the variation in microclimatic  
11 variables, and the primary variable (>36%) for PC1 was mean annual soil temperature (Figure  
12 S1). Therefore, besides SMI, mean annual soil temperature was included in the models.

13 Lastly, we investigated the relationship between trait means and CV to test whether  
14 CV varies independently from trait means. At population and genet level, we ran LMMs with  
15 the trait mean as the response variable, CV and quadratic CV of the trait as fixed effects, and  
16 nested hierarchical levels as random effect. At clone and ramet level, shade and drought  
17 treatments, along with all possible interactions with CV and quadratic CV of the traits  
18 included as fixed effects, with region, population, and genet as nested random effect for clone  
19 level, and the fully nested hierarchical levels of organisation (*i.e.*, region, population, genet,  
20 and clone) as nested random effect for ramet level.

21 All statistical analyses were conducted with R version 4.1.2 (R Core Team, 2021).  
22 Shapiro-Wilk and Bartlett tests were used to assess whether the assumptions of normality and  
23 homoscedasticity of model residuals were met, respectively. When at least one of these  
24 assumptions was violated, the response variable was transformed. Stepwise backward model  
25 selection was applied in all models (Pearce and Ferrier, 2000).

1

## 2 **Results**

3 We measured a total of 2,304 leaves, from a total of 517 ramets, arising from 135 clones,  
4 originating from 71 genets, sampled from 21 different populations across the three different  
5 regions (ALB, HAI, and SCH; Table S1). Height was measured on all 517 ramets.

### 6 *Variance partitioning on hierarchical levels of organisation*

7 Partitions of trait variation into hierarchically nested levels of organisation revealed that the  
8 variation within ramets ( $CV_{IR}$ ) was the main source of variation in leaf width and length,  
9 accounting for 92.5% and 85.8% of the total variation, respectively (Fig. 2). For leaf width,  
10 7.5% was explained by the variation at intra-clone level, whereas for leaf length the remaining  
11 variation, 11.2%, was mainly explained by the variation at intra-clone level, and 2.8% by  
12 population, and the remaining 0.2% was unexplained variation (Fig. 2). Variation in height  
13 was primarily explained by clone level with 71.4%, genet level explained 17.9%, population  
14 explained 3.8%, and the remaining 6.9% was unexplained variation (Fig. 2).

### 15 *Genetic basis of CV*

16 The hierarchical level of organisation with the best goodness of fit, *i.e.*, that best explained  
17 CV, varied among traits. At the intra-genet level (reflecting a combination of genetic and  
18 plastic effects), AIC values showed that region best explained  $CV_{IG}$  in leaf width, and  
19 population best explained  $CV_{IG}$  in plant height, whereas no differences ( $\geq 2$ ) between AIC in  
20 models were found for  $CV_{IG}$  in leaf length (Table 1). At the intra-clone level (reflecting  
21 genetic effects), none of the hierarchical levels of organisation had an effect on the  $CV_{IC}$  in  
22 height, leaf length, or leaf width (Table 1). Finally, on the intra-ramet level (reflecting genetic  
23 effects), region had the lowest AIC values, and therefore best explained  $CV_{IR}$ , in leaf width,  
24 whereas population and genet best explained  $CV_{IR}$  in leaf length.

1 *Microclimate and treatment effects on CV*

2  $CV_{IP}$ ,  $CV_{IG}$ , and  $CV_{IC}$  in height were all significantly affected by mean annual soil  
3 temperature (Table 2): CV decreased with increasing mean annual soil temperature (Fig. 3).  
4 Additionally,  $CV_{IP}$  in leaf length was also significantly negatively affected by mean annual  
5 soil temperature. No significant effect of mean annual soil temperature was found on CV in  
6 leaf width. No significant effect of SMI was found on CV in any of the vegetative traits.

7         The early shading treatment significantly increased  $CV_{IC}$  in leaf length compared to  
8 control shading (Table 2; Fig. 4A). The two-way interaction between shade and drought was  
9 significant for  $CV_{IR}$  in leaf width (Table 2), with the combination of drought treatment and  
10 control shading resulting in lower  $CV_{IR}$  compared to the other treatment combinations (Fig.  
11 4B). No significant treatment effects were found in CV in height in any of the hierarchical  
12 levels (Table 2).

13 *Trait mean and CV relationship*

14 Mean height at intra-genet and intra-clone level was significantly affected by the quadratic  
15 terms of  $CV_{IG}$  and  $CV_{IC}$  in height (Table 3). Mean leaf length was significantly affected by  
16 the quadratic terms of  $CV_{IG}$ ,  $CV_{IC}$ , and  $CV_{IR}$ . Mean leaf width was significantly affected by  
17 the  $CV_{IG}$ ,  $CV_{IC}$ , and  $CV_{IR}$  of leaf width. Furthermore, our results show a significant  
18 interaction for mean leaf length on the intra-clone level between quadratic  $CV_{IC}$  and drought  
19 treatment (Table 3). Specifically, mean leaf length and quadratic  $CV_{IC}$  of leaf length are  
20 related under the drought treatment, but they are independent under the control treatment (Fig.  
21 5).

22

23 **Discussion**

1 This experimental study using *G. odoratum* as model species revealed (1) that intra-ramet  
2 variation, which is the lowest hierarchical level in our system, explains the vast majority of  
3 overall leaf trait variation in the populations, followed by intra-clone variation. These two  
4 levels represent intra-individual variation, and our results thus confirm that intra-individual  
5 variation can exceed inter-individual variation (*i.e.*, intra-population and inter-genet  
6 variation), as previously observed (Herrera 2017); (2) that intra-specific and intra-individual  
7 trait variation at different scales is partly genetically based and (3) that this variation may  
8 have been the result of selection by microclimatic conditions in the populations of origin.  
9 Furthermore, (4) intra-ramet and intra-clone variation in leaf traits vary under induced drought  
10 and early shading, possibly as a functional response to changes in water and light conditions.  
11 Finally, (5) drought also led to a dependent relationship between mean leaf size and intra-  
12 individual variation in leaf size, as a plastic response that may promote population stability in  
13 stressful conditions (Nicotra *et al.*, 2010).

#### 14 *Variance partitioning across hierarchical levels of biological organisation*

15 Our variance partitioning analysis showed that the lowest hierarchical level of organisation,  
16 intra-ramet variation, explained the majority of total variation in leaf length and leaf width,  
17 followed by intra-clone variation. Additionally, variation in plant height was primarily  
18 explained by the intra-clone level. This is congruent with what has been observed in previous  
19 studies, where intra-ramet (or intra-individual variation) also explained more variation in leaf  
20 traits, fruits and seeds than the variation among plants (Herrera *et al.*, 2015; Herrera, 2017).  
21 These striking patterns in the two lowest hierarchical levels could be due to leaf traits changes  
22 through time while the plant develops, whereas at higher hierarchical levels the trait  
23 differences are already summarised across ramets and therefore average out the intra-ramet  
24 variation (*i.e.*, leaves on different ramets are on average more similar to each other than leaves  
25 within a ramet). Taken together, these results support the notion that intra-individual variation

1 is responsible for the larger share of the total intra-specific variation, potentially driving  
2 population performance and plastic responses to environmental changes. The topic of intra-  
3 individual variation is thus worthy of future investigation, and our results justify further  
4 research on a deeper mechanistic explanation for intra-individual variation and how these  
5 might enhance population variation and persistence.

#### 6 *Genetic differentiation in intra-individual variation*

7 In our common garden experiment, any differences in intra-clone and intra-ramet CV among  
8 regions, populations, or genets of origin suggest the existence of genetically based variation in  
9 intra-individual variation. We revealed a general pattern of the highest hierarchical levels of  
10 region and population having the best goodness of fit, thereby being the primary factors  
11 explaining genetic variation in CV at the lower hierarchical levels of intra-genet and intra-  
12 ramet. This pattern could well reflect the strong versatility of clonal plants, allowing  
13 individuals to adjust their life-history in terms of vegetative reproduction, the placement of  
14 ramets and the degree of integration between ramets according to the local environmental  
15 conditions (Fischer and van Kleunen, 2001). An individual from a fine-grained heterogeneous  
16 environment and microclimate is more likely to invest in increased variation in order to deal  
17 with environmental changes and predictability on a small spatial scale, compared to an  
18 individual from a coarse-grained heterogeneous or even homogenous environment (Winn,  
19 1996b; Stark *et al.*, 2017). The observed genetic variation in the CV of plant height, leaf  
20 length and width, primarily explained by region and population, may very well have been  
21 driven by the different environments among regions (ALB, HAI, and SCH) and among all  
22 sampled populations (see next section).

#### 23 *Effects of microclimate of origin and treatments on CV*

24 Mean annual soil temperature from the populations of origin was found to have a significant  
25 effect on variation in height at all hierarchical levels of organisation, as well as on variation in

1 leaf length at intra-population level. The negative association between CV in height and mean  
2 annual soil temperature indicates that individuals originating from colder sites show increased  
3 variation, in support of a stress-induced variability hypothesis, stating that abiotic stress  
4 increases intra-specific variation (Kuppler *et al.*, 2020). Colder plots from the Biodiversity  
5 Exploratories are primarily within coniferous forests, resulting in more or less constant shaded  
6 conditions (Willems *et al.*, 2021). Therefore, a possible explanation for the observed  
7 increased height variation in individuals originating from colder plots could be, that these  
8 individuals have evolved an optimized way to capture light under light-limited conditions by  
9 increasing their variation, potentially by number of leaves and the positioning of them  
10 (Valladares and Brites, 2004). This relationship thus suggest that intra-individual variation  
11 may respond adaptively to various environmental conditions (Herrera *et al.*, 2015) may and  
12 ultimately lead to adaptation to long-term microclimatic conditions (May *et al.*, 2017).

13         At the intra-clone level, intra-individual variation in leaf length was found to increase  
14 under two-weeks earlier shading, indicating a rapid phenotypic change within clonal plants in  
15 response to environmental drivers. Under shaded conditions, common phenotypic plastic  
16 responses are an increased mean leaf size and specific leaf area (Pigliucci and Kolodynska,  
17 2002; Sultan, 2003). After shading was applied, leaf length values may have adjusted in  
18 emerging whorls on the developing ramets, causing the intra-individual variation to increase  
19 under the early shading treatment. However, at the intra-ramet level the interaction between  
20 the shading and drought treatment had a significant effect on the intra-individual variation in  
21 leaf width: compared to control shading without drought, early shading with drought, and  
22 early shading without drought treatment combinations, the intra-individual variation was  
23 considerably lower in individuals exposed to the control shading with drought treatment  
24 combination. Possibly, the increase in intra-individual variation observed in the early shading  
25 with drought treatment could be a result of plant responses in various directions in response to



1 drastic environmental conditions (Winn, 1996b; Sun *et al.*, 2021), potentially benefitting the  
2 individual performance under the drought conditions, perhaps through facilitating drought  
3 tolerance. Another strategy to deal with drought is to avoid unnecessary evapotranspiration by  
4 minimizing leaf area (Marron *et al.*, 2003; Liu and Stützel, 2004; Wellstein *et al.*, 2017).  
5 Newly developing leaves have the opportunity to respond to current environmental  
6 conditions, given that there is sufficient phenotypic plasticity for rapid adjustments. However,  
7 the low intra-ramet variation in leaf width observed under the drought treatment implies that  
8 the individuals did not respond to the drought conditions by further minimizing their leaf area,  
9 leading to the low intra-individual variation.

10         With regard to intra-individual responses, it is important to consider the morphology  
11 of the study species. *Galium odoratum* has its leaves positioned in multiple whorls on a single  
12 ramet. These whorls and leaves generally increase in size with increasing distance from the  
13 ground surface, with the exception of the top whorl which is typically still under  
14 development. Therefore, some variation would always be present, even under a hypothetically  
15 constant environment, simply due to the structure and architecture of the leaves and whorls  
16 (Fischer and van Kleunen, 2001). Nevertheless, this inherent intra-individual variation could  
17 very well be a result of evolution of *G. odoratum*, for instance to maximize light capture  
18 (Chazdon and Kaufmann, 1993; Reich *et al.*, 2003).

#### 19 *Relationship between trait means and coefficient of variation*

20 Significant relationships between trait means and CV would suggest that these two  
21 components are interdependent and that one could drive the other. However, if means and CV  
22 are independent, intra-individual variation should be considered as an individual property in  
23 itself (Herrera, 2017), able to respond and evolve unconstrained with respect to mean trait  
24 values. We observed both linear and quadratic relationships between trait means and the CV  
25 on intra-genet, intra-clone and intra-ramet level, but also in some cases an absence of any

1 relationship. The quadratic functions suggest a more complex relationship, where  
2 substantially low and high intra-individual variation values reduce mean plant responses while  
3 intermediate intra-individual variation values could enhance a plant's performance, in line  
4 with stabilizing selective regimes found in previous studies (March-Salas et al. 2021). In this  
5 sense, the optimal strategy for some plants could be to invest in intermediate intra-individual  
6 variation.

7         At the intra-clone level, we found that the relationship between mean leaf length and  
8 leaf length CV depends on the drought treatment: Under the control conditions, there were no  
9 relationships between mean leaf length and its variation, whereas under drought conditions  
10 there existed a slight concave relationship that shows a decreasing mean leaf length with  
11 increasing intra-individual variation. Thus, under drought stress, plants with decreasing leaf  
12 length become less uniform in their leaf traits. This difference between treatment responses  
13 indicates that drought conditions affect the relationship between trait means and their  
14 variation. In this sense, it could be that plants promote intra-individual changes to optimise  
15 final plant trait values, or potentially facilitate eco-physiological traits through water use  
16 efficiency and decreased evapotranspiration under stressful conditions (Marron *et al.*, 2003;  
17 Liu and Stützel, 2004; Wellstein *et al.*, 2017). Smaller plants may therefore be particularly  
18 prone to increase their intra-individual variation in order to adjust to the stressful drought  
19 conditions due to their smaller leaf sizes.

20

## 21 **Conclusion**

22 In sum, using a common-garden experiment we demonstrated that intra-individual variation  
23 introduces an important component of variation in populations. Intra-ramet variation  
24 explained the vast majority of total population variation, suggesting a relevant role of  
25 phenotypic variation within individuals for adjustments to the local environmental conditions

1 and ultimately for population performance. Furthermore, our results indicate that intra-  
2 individual variation in vegetative traits may be partially genetically based, indicating that it  
3 could be under the influences of neutral or selective evolutionary processes. Relationships  
4 between mean annual soil temperature at the sites of origin and intra-individual variation in  
5 vegetative traits suggest that the environment shaped genetically based among-population  
6 variation in intra-individual variation. We also showed that intra-individual variation responds  
7 plastically to drought and shading treatments, which may allow plants to maintain or improve  
8 overall performance under stressful or varying environmental conditions. However, we lack  
9 strong support for mechanistic explanations for most patterns in intra-individual variation,  
10 potentially mediated by eco-physiological and epigenetic modifications. We therefore  
11 conclude that, overall, the observed patterns in intra-individual variation suggest its important  
12 role for genetic adaptation to microclimatic variation among populations and for plastic  
13 adjustments to rapidly changing microclimatic conditions within populations, respectively.

14

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11

## 12 **Conflict of interest**

13 The authors have no conflict of interest to declare.

14

## 15 **Author contributions**

16 CM, MMS, PDF and JFS designed the study. CM conducted the field sampling and the  
17 common garden experiment. CM, MMS, and JFS analysed the data with input from PDF and  
18 JK. CM wrote the first draft of the manuscript with all co-authors contributing to revisions.

## 19 **References**

- 20 **Albert CH, Thuiller W, Yoccoz NG, et al. 2010.** Intraspecific functional variability: extent, structure  
21 and sources of variation. *Journal of Ecology* **98**: 604–613. doi:10.1111/j.1365-2745.2010.01651.x.
- 22 **Alonso C, Pérez R, Bazaga P, Medrano M, Herrera CM. 2018.** Within-plant variation in seed size  
23 and inflorescence fecundity is associated with epigenetic mosaicism in the shrub *Lavandula*  
24 *latifolia* (Lamiaceae). *Ann Bot* **121**: 153–160. doi:10.1093/aob/mcx140.
- 25 **Bates D, Sarkar D, Bates MD, Matrix L. 2007.** *The lme4 package: R package version.*

- 1 **Büntgen U, Urban O, Krusic PJ, et al. 2021.** Recent European drought extremes beyond Common  
2 Era background variability. *Nat. Geosci.* **14**: 190–196. doi:10.1038/s41561-021-00698-0.
- 3 **Chazdon RL, Kaufmann S. 1993.** Plasticity of Leaf Anatomy of Two Rain Forest Shrubs in Relation  
4 to Photosynthetic Light Acclimation. *Funct Ecology* **7**: 385. doi:10.2307/2390025.
- 5 **Chen J, Dai A, Zhang Y. 2019.** Projected Changes in Daily Variability and Seasonal Cycle of Near-  
6 Surface Air Temperature over the Globe during the Twenty-First Century. *Journal of Climate* **32**:  
7 8537–8561. doi:10.1175/JCLI-D-19-0438.1.
- 8 **Dai A. 2013.** Increasing drought under global warming in observations and models. *Nature Clim*  
9 *Change* **3**: 52–58. doi:10.1038/nclimate1633.
- 10 **Dai A, Zhao T, Chen J. 2018.** Climate Change and Drought: a Precipitation and Evaporation  
11 Perspective. *Curr Clim Change Rep* **4**: 301–312. doi:10.1007/s40641-018-0101-6.
- 12 **De Frenne P, Graae BJ, Rodríguez-Sánchez F, et al. 2013.** Latitudinal gradients as natural  
13 laboratories to infer species' responses to temperature. *Journal of Ecology* **101**: 784–795.  
14 doi:10.1111/1365-2745.12074.
- 15 **Des Roches S, Post DM, Turley NE, et al. 2018.** The ecological importance of intraspecific variation.  
16 *Nat Ecol Evol* **2**: 57–64. doi:10.1038/s41559-017-0402-5.
- 17 **Esau K. 1960.** Anatomy of Seed Plants. *Soil Science* **90**: 149.
- 18 **Fang Y, Xiong L. 2015.** General mechanisms of drought response and their application in drought  
19 resistance improvement in plants. *Cell. Mol. Life Sci.* **72**: 673–689. doi:10.1007/s00018-014-1767-  
20 0.
- 21 **Fischer M, Bossdorf O, Gockel S, et al. 2010.** Implementing large-scale and long-term functional  
22 biodiversity research: The Biodiversity Exploratories. *Basic and Applied Ecology* **11**: 473–485.  
23 doi:10.1016/j.baae.2010.07.009.
- 24 **Fischer M, van Kleunen M. 2001.** On the evolution of clonal plant life histories. *Evolutionary*  
25 *Ecology* **15**: 565–582. doi:10.1023/A:1016013721469.
- 26 **Forsman A. 2015.** Rethinking phenotypic plasticity and its consequences for individuals, populations  
27 and species. *Heredity* **115**: 276–284. doi:10.1038/hdy.2014.92.

1 **Fox J, Weisberg S, Adler D, et al. 2012.** *Package 'car'*: Vienna: R Foundation for Statistical  
2 Computing.

3 **Franks SJ, Sim S, Weis AE. 2007.** Rapid evolution of flowering time by an annual plant in response  
4 to a climate fluctuation. *Proc. Natl. Acad. Sci. U.S.A.* **104**: 1278–1282.  
5 doi:10.1073/pnas.0608379104.

6 **Frederiksen S, Rasmussen FN. 2006.** *Dansk flora*: Gyldendal A/S.

7 **Govaert L, Pantel JH, Meester L de. 2016.** Eco-evolutionary partitioning metrics: assessing the  
8 importance of ecological and evolutionary contributions to population and community change.  
9 *Ecology letters* **19**: 839–853. doi:10.1111/ele.12632.

10 **Heberling JM, McDonough MacKenzie C, Fridley JD, Kalisz S, Primack RB. 2019.** Phenological  
11 mismatch with trees reduces wildflower carbon budgets. *Ecology letters* **22**: 616–623.  
12 doi:10.1111/ele.13224.

13 **Hedrick PW. 2009.** *Genetics of Populations*: Jones & Bartlett Publishers.

14 **Herrera CM. 2009.** *Multiplicity in Unity*: University of Chicago Press.

15 **Herrera CM. 2017.** The ecology of subindividual variability in plants: patterns, processes, and  
16 prospects. *Web Ecol.* **17**: 51–64. doi:10.5194/we-17-51-2017.

17 **Herrera CM, Medrano M, Bazaga P. 2015.** Continuous within-plant variation as a source of  
18 intraspecific functional diversity: Patterns, magnitude, and genetic correlates of leaf variability in  
19 *Helleborus foetidus* (Ranunculaceae). *American Journal of Botany* **102**: 225–232.  
20 doi:10.3732/ajb.1400437.

21 **Kinnison MT, Hairston NG. 2007.** Eco-evolutionary conservation biology: contemporary evolution  
22 and the dynamics of persistence. *Funct Ecology* **21**: 444–454. doi:10.1111/j.1365-  
23 2435.2007.01278.x.

24 **Kuppler J, Albert CH, Ames GM, et al. 2020.** Global gradients in intraspecific variation in  
25 vegetative and floral traits are partially associated with climate and species richness. *Global*  
26 *Ecology and Biogeography* **29**: 992–1007. doi:10.1111/geb.13077.

27 **Lenth RV. 2016.** Least-Squares Means: The R Package lsmeans. *J. Stat. Soft.* **69**: 1–33.  
28 doi:10.18637/jss.v069.i01.

- 1 **Liu F, Stützel H. 2004.** Biomass partitioning, specific leaf area, and water use efficiency of vegetable  
2 amaranth (*Amaranthus* spp.) in response to drought stress. *Scientia Horticulturae* **102**: 15–27.  
3 doi:10.1016/j.scienta.2003.11.014.
- 4 **March-Salas M, Fandos G, Fitze PS. 2021.** Effects of intrinsic environmental predictability on intra-  
5 individual and intra-population variability of plant reproductive traits and eco-evolutionary  
6 consequences. *Ann Bot* **127**: 413–423. doi:10.1093/aob/mcaa096.
- 7 **Marron N, Dreyer E, Boudouresque E, et al. 2003.** Impact of successive drought and re-watering  
8 cycles on growth and specific leaf area of two *Populus x canadensis* (Moench) clones, 'Dorskamp'  
9 and 'Luisa\_Avanzo'. *Tree Physiol* **23**: 1225–1235. doi:10.1093/treephys/23.18.1225.
- 10 **Martinez KA, Fridley JD. 2018.** Acclimation of leaf traits in seasonal light environments: Are non-  
11 native species more plastic? *J Ecol* **106**: 2019–2030. doi:10.1111/1365-2745.12952.
- 12 **May R-L, Warner S, Wingler A. 2017.** Classification of intra-specific variation in plant functional  
13 strategies reveals adaptation to climate. *Ann Bot* **119**: 1343–1352. doi:10.1093/aob/mcx031.
- 14 **Millar CI, Stephenson NL. 2015.** Temperate forest health in an era of emerging megadisturbance.  
15 *Science* **349**: 823–826. doi:10.1126/science.aaa9933.
- 16 **Nicotra AB, Atkin OK, Bonser SP, et al. 2010.** Plant phenotypic plasticity in a changing climate.  
17 *Trends in Plant Science* **15**: 684–692. doi:10.1016/j.tplants.2010.09.008.
- 18 **Pearce J, Ferrier S. 2000.** An evaluation of alternative algorithms for fitting species distribution  
19 models using logistic regression. *Ecological Modelling* **128**: 127–147. doi:10.1016/S0304-  
20 3800(99)00227-6.
- 21 **Pigliucci M, Kolodynska A. 2002.** Phenotypic plasticity to light intensity in *Arabidopsis thaliana*:  
22 invariance of reaction norms and phenotypic integration. *Evolutionary Ecology* **16**: 27–47.  
23 doi:10.1023/A:1016073525567.
- 24 **R Core Team. 2021.** *R: A language and environment for statistical computing*. R Foundation for  
25 Statistical Computing, Vienna, Austria.
- 26 **Reich PB, Wright IJ, Cavender-Bares J, et al. 2003.** The Evolution of Plant Functional Variation:  
27 Traits, Spectra, and Strategies. *International Journal of Plant Sciences* **164**: S143-S164.  
28 doi:10.1086/374368.

- 1 **Schall P, Ammer C. 2013.** How to quantify forest management intensity in Central European forests.  
2 *Eur J Forest Res* **132**: 379–396. doi:10.1007/s10342-013-0681-6.
- 3 **Schall P, Ammer C. 2014.** *SMI - Silvicultural management intensity index on all forest EPs, 2008-*  
4 *2014, Version 3 Dataset ID = 17746*: Biodiversity Exploratories Information System.  
5 <https://www.bexis.uni-jena.de/>.
- 6 **Scheepens JF, Frei ES, Stöcklin J. 2013.** Glacial history affected phenotypic differentiation in the  
7 alpine plant, *Campanula thyrsoidea*. *PLOS ONE* **8**: e73854. doi:10.1371/journal.pone.0073854.
- 8 **Siefert A, Violle C, Chalmandrier L, et al. 2015.** A global meta-analysis of the relative extent of  
9 intraspecific trait variation in plant communities. *Ecology letters* **18**: 1406–1419.  
10 doi:10.1111/ele.12508.
- 11 **Sobral M, Guitián J, Guitián P, Larrinaga AR. 2013.** Selective pressure along a latitudinal gradient  
12 affects subindividual variation in plants. *PLOS ONE* **8**: e74356.  
13 doi:10.1371/journal.pone.0074356.
- 14 **Stark J, Lehman R, Crawford L, Enquist BJ, Blonder B. 2017.** Does environmental heterogeneity  
15 drive functional trait variation? A test in montane and alpine meadows. *Oikos* **126**: 1650–1659.  
16 doi:10.1111/oik.04311.
- 17 **Sultan SE. 2003.** Phenotypic plasticity in plants: a case study in ecological development. *Evol Dev* **5**:  
18 25–33. doi:10.1046/j.1525-142x.2003.03005.x.
- 19 **Sun C, Ali K, Yan K, et al. 2021.** Exploration of Epigenetics for Improvement of Drought and Other  
20 Stress Resistance in Crops: A Review. *Plants* **10**: 1226. doi:10.3390/plants10061226.
- 21 **Taudiere A, Violle C. 2016.** *cati*: an R package using functional traits to detect and quantify multi-  
22 level community assembly processes. *Ecography* **39**: 699–708. doi:10.1111/ecog.01433.
- 23 **Trenberth KE, Dai A, van der Schrier G, et al. 2014.** Global warming and changes in drought.  
24 *Nature Clim Change* **4**: 17–22. doi:10.1038/nclimate2067.
- 25 **Turesson G. 1922.** The genotypical response of the plant species to the habitat. *Hereditas* **3**: 211–350.  
26 doi:10.1111/j.1601-5223.1922.tb02734.x.
- 27 **Valladares F, Brites D. 2004.** Leaf phyllotaxis: Does it really affect light capture? *Plant Ecology* **174**:  
28 11–17. doi:10.1023/B:VEGE.0000046053.23576.6b.



- 1 **Violle C, Enquist BJ, McGill BJ, et al. 2012.** The return of the variance: intraspecific variability in  
2 community ecology. *Trends in Ecology & Evolution* **27**: 244–252. doi:10.1016/j.tree.2011.11.014.
- 3 **Vitasse Y, Baumgarten F, Zohner CM, et al.** The great acceleration of plant phenological shifts.  
4 *Nature Clim Change*, in press. doi:10.1038/s41558-022-01283-y.
- 5 **Vrieze SI. 2012.** Model selection and psychological theory: a discussion of the differences between  
6 the Akaike information criterion (AIC) and the Bayesian information criterion (BIC).  
7 *Psychological Methods* **17**: 228–243. doi:10.1037/a0027127.
- 8 **Wellstein C, Poschlod P, Gohlke A, et al. 2017.** Effects of extreme drought on specific leaf area of  
9 grassland species: A meta-analysis of experimental studies in temperate and sub-Mediterranean  
10 systems. *Global Change Biology* **23**: 2473–2481. doi:10.1111/gcb.13662.
- 11 **West-Eberhard MJ. 2003.** *Developmental Plasticity and Evolution*: Oxford University Press.
- 12 **Westerband AC, Funk JL, Barton KE. 2021.** Intraspecific trait variation in plants: a renewed focus  
13 on its role in ecological processes. *Ann Bot* **127**: 397–410. doi:10.1093/aob/mcab011.
- 14 **Willems FM, Scheepens JF, Ammer C, et al. 2021.** Spring understory herbs flower later in  
15 intensively managed forests. *Ecological applications a publication of the Ecological Society of*  
16 *America* **31**: e02332. doi:10.1002/eap.2332.
- 17 **Winn AA. 1996a.** Adaptation to fine-grained environmental variation: An analysis of within-  
18 individual leaf variation in an annual plant. *Evolution* **50**: 1111–1118. doi:10.1111/j.1558-  
19 5646.1996.tb02351.x.
- 20 **Winn AA. 1996b.** The contributions of programmed developmental change and phenotypic plasticity  
21 to within-individual variation in leaf traits in *Dicerandra linearifolia*. *J Evolution Biol* **9**: 737–752.  
22 doi:10.1046/j.1420-9101.1996.9060737.x.

23  
24

25 **Table 1.** Comparison of models for genetic variation in CV of height, leaf length and width in *Galium odoratum* at different hierarchical levels. Hierarchical  
 26 levels were added one at a time as nested random factor, and goodness of fit was compared using Akaike Information Criterion (AIC). Significantly lower AIC  
 27 values ( $\Delta AIC \geq 2$ ) that explain the best goodness of fit are indicated in bold.

<b>Akaike Information Criterion (AIC) values for genetic basis of trait CV</b>			
	<b>Region</b>	<b>Region/Population</b>	<b>Region/Population/Genet</b>
<b>INTRA-GENET (CV<sub>IG</sub>, N = 71)</b>			
Height	- 123.6	- <b>126.4</b>	
Leaf length	- 137.8	- 136.7	
Leaf width	- <b>165.3</b>	- 163.3	
<b>INTRA-CLONE (CV<sub>IC</sub>, N = 135)</b>			
Height	- 183.6	- 183.7	- 184.2
Leaf length	- 204.1	- 205.8	- 203.8
Leaf width	- 251.8	- 250.9	- 248.9
<b>INTRA-RAMET (CV<sub>IR</sub>, N = 517)</b>			
Leaf length	- 545.6	- 549.8	- 549.5
Leaf width	- <b>609.2</b>	- 607.2	- 605.9

28

29 **Table 2.** Results of mixed-effects models of CV in height, leaf length and leaf width in *Galium odoratum* at different hierarchical levels explained by  
 30 environmental variables and experimental treatments. Stepwise backwards selection was applied to obtain the minimum adequate models. Parameters used in  
 31 the minimum adequate model are indicated with bold. Marginally significant effects ( $0.1 > P > 0.05$ ) are shown in italics. If stepwise backwards selection did  
 32 not lead to removing model parameters, estimates from the full model are shown. Parameter estimates (PE) for continuous variables are indicated with  $\uparrow$  for  
 33 positive and  $\downarrow$  for negative. Chi-square ( $X^2$ ) test values, degrees of freedom (df) and P-values (P) are given.

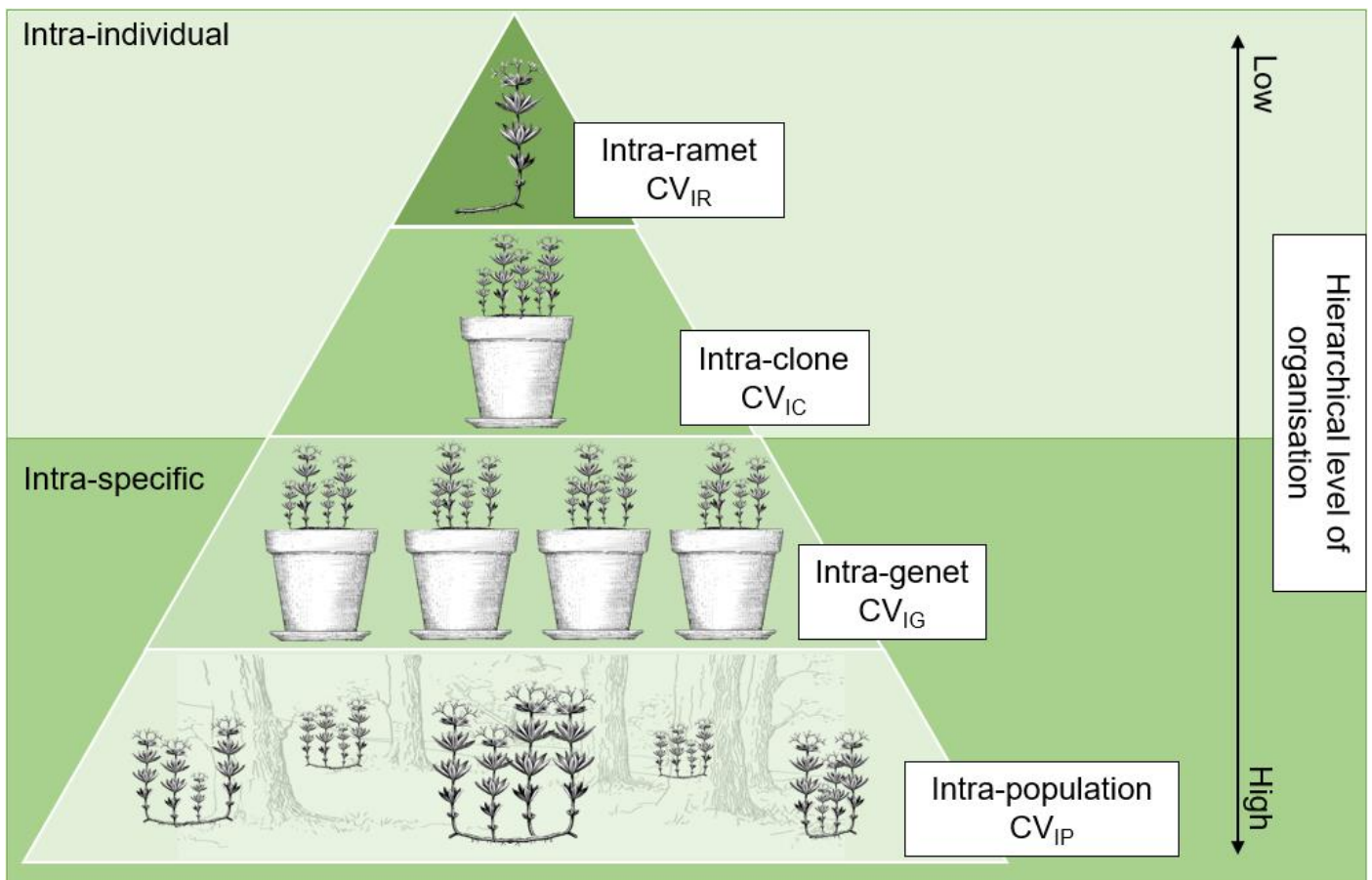
<b>INTRA-POPULATION (CV<sub>IP</sub>, N = 22)</b>												
<i>Predictors</i>	<b>Height</b>				<b>Leaf length</b>				<b>Leaf width</b>			
	PE	$X^2$	df	P	PE	$X^2$	df	P	PE	$X^2$	df	P
Soil temperature	$\downarrow$	<b>5.237</b>	<b>1</b>	<b>0.022 *</b>	$\downarrow$	<b>3.870</b>	<b>1</b>	<b>0.049 *</b>	$\downarrow$	<i>2.844</i>	<i>1</i>	<i>0.092</i>
SMI	$\uparrow$	0.546	1	0.460	$\downarrow$	0.036	1	0.850	$\uparrow$	0.054	1	0.817
<b>INTRA-GENET (CV<sub>IG</sub>, N = 71)</b>												
<i>Predictors</i>	<b>Height</b>				<b>Leaf length</b>				<b>Leaf width</b>			
	PE	$X^2$	df	P	PE	$X^2$	df	P	PE	$X^2$	df	P
Soil temperature	$\downarrow$	<b>7.682</b>	<b>1</b>	<b>0.006 **</b>	$\downarrow$	1.012	1	0.314	$\downarrow$	0.130	1	0.718
SMI	$\downarrow$	0.045	1	0.831	$\uparrow$	0.482	1	0.488	$\uparrow$	1.616	1	0.204
<b>INTRA-CLONE (CV<sub>IC</sub>, N = 135)</b>												
<i>Predictors</i>	<b>Height</b>				<b>Leaf length</b>				<b>Leaf width</b>			
	PE	$X^2$	df	P	PE	$X^2$	df	P	PE	$X^2$	df	P
Soil temperature	$\downarrow$	<b>8.686</b>	<b>1</b>	<b>0.003 **</b>	$\downarrow$	0.658	1	0.417	$\downarrow$	0.050	1	0.823
SMI	$\downarrow$	0.077	1	0.782	$\uparrow$	0.002	1	0.965	$\uparrow$	0.477	1	0.490
Shade (S)		2.250	1	0.134		<b>4.299</b>	<b>1</b>	<b>0.038 *</b>		<i>3.127</i>	<i>1</i>	<i>0.077</i>
Drought (D)		0.046	1	0.831		0.662	1	0.416		1.407	1	0.236
S $\times$ D		1.813	2	0.404		0.004	1	0.948		0.510	1	0.475
<b>INTRA-RAMET (CV<sub>IR</sub>, N = 517)</b>												
<i>Predictors</i>	<b>Leaf length</b>				<b>Leaf width</b>							
	PE	$X^2$	df	P	PE	$X^2$	df	P				
Soil temperature	$\downarrow$	0.799	1	0.371	$\downarrow$	0.869	1	0.351				
SMI	$\downarrow$	0.308	1	0.579	$\downarrow$	0.031	1	0.861				
Shade (S)		2.592	1	0.107		<b>2.348</b>	<b>1</b>	<b>0.125</b>				
Drought (D)		1.254	1	0.263		<b>1.508</b>	<b>1</b>	<b>0.220</b>				
S $\times$ D		3.219	1	0.073		<b>6.229</b>	<b>1</b>	<b>0.012 *</b>				

34 **Table 3.** The full models of all the predictors for trait means. Stepwise backwards selection was applied to obtain the minimum adequate models. Parameters  
 35 used in the final model are indicated with bold. Marginal effects ( $0.1 > P > 0.05$ ) are shown in italics. If the stepwise backwards selection did not lead to  
 36 removed parameters in the model, estimates from the full model are shown.  $X^2$  values, degrees of freedom (df) and P-values (P) are given. No significant  
 37 results were found for intra-population ( $CV_{IP}$ ), and thus this level is not shown.

<b>INTRA-GENET (<math>CV_{IG}</math>, N = 71)</b>									
<i>Predictors</i>	<b>Height</b>			<b>Leaf length</b>			<b>Leaf width</b>		
	$X^2$	df	P	$X^2$	df	P	$X^2$	df	P
Trait CV	<i>3.453</i>	<i>1</i>	<i>0.063</i>	0.604	1	0.437	<b>7.395</b>	<b>1</b>	<b>0.007 **</b>
Trait CV <sup>2</sup>	<b>4.178</b>	<b>1</b>	<b>0.041 *</b>	<b>6.386</b>	<b>1</b>	<b>0.012 *</b>	0.023	1	0.880
<b>INTRA-CLONE (<math>CV_{IC}</math>, N = 135)</b>									
<i>Predictors</i>	<b>Height</b>			<b>Leaf length</b>			<b>Leaf width</b>		
	$X^2$	df	P	$X^2$	df	P	$X^2$	df	P
Trait CV	2.482	1	0.115	1.173	1	0.279	<b>20.149</b>	<b>1</b>	<b>&lt;0.001 ***</b>
Trait CV <sup>2</sup>	<b>3.951</b>	<b>1</b>	<b>0.047 *</b>	<b>10.965</b>	<b>1</b>	<b>&lt;0.001 ***</b>	0.006	1	0.937
Shade (S)	0.430	1	0.512	0.790	1	0.374	0.359	1	0.549
Drought (D)	0.002	1	0.962	<b>1.656</b>	<b>1</b>	<b>0.198</b>	1.137	1	0.286
Trait CV × S	0.258	1	0.612	0.010	1	0.929	0.098	1	0.754
Trait CV × D	0.363	1	0.547	0.121	1	0.728	0.080	1	0.777
Trait CV <sup>2</sup> × S	0.386	1	0.534	0.551	1	0.458	0.174	2	0.917
Trait CV <sup>2</sup> × D	0.081	1	0.776	<b>4.705</b>	<b>1</b>	<b>0.030 *</b>	0.224	2	0.894
Trait CV × S × D	0.338	1	0.561	0.068	1	0.794	0.515	2	0.773
Trait CV <sup>2</sup> × S × D	1.937	3	0.586	2.611	2	0.271	1.053	4	0.902
<b>INTRA-RAMET (<math>CV_{IR}</math>, N = 517)</b>									
<i>Predictors</i>	<b>Leaf length</b>			<b>Leaf width</b>					
	$X^2$	df	P	$X^2$	df	P			
Trait CV	0.087	1	0.768	<b>112.22</b>	<b>1</b>	<b>&lt; 0.001 ***</b>			
Trait CV <sup>2</sup>	<b>67.162</b>	<b>1</b>	<b>&lt; 0.001 ***</b>	0.760	1	0.384			
Shade (S)	0.434	1	0.510	0.646	1	0.422			
Drought (D)	<i>3.071</i>	<i>1</i>	<i>0.080</i>	1.995	1	0.158			
Trait CV × S	0.002	1	0.965	0.001	1	0.970			
Trait CV × D	0.521	1	0.771	0.014	1	0.907			
Trait CV <sup>2</sup> × S	0.215	1	0.643	0.331	1	0.565			

Trait $CV^2 \times D$	0.004	1	0.950	1.524	2	0.469
Trait $CV \times S \times D$	0.036	2	0.982	0.050	2	0.975
Trait $CV^2 \times S \times D$	0.533	2	0.766	0.050	1	0.823

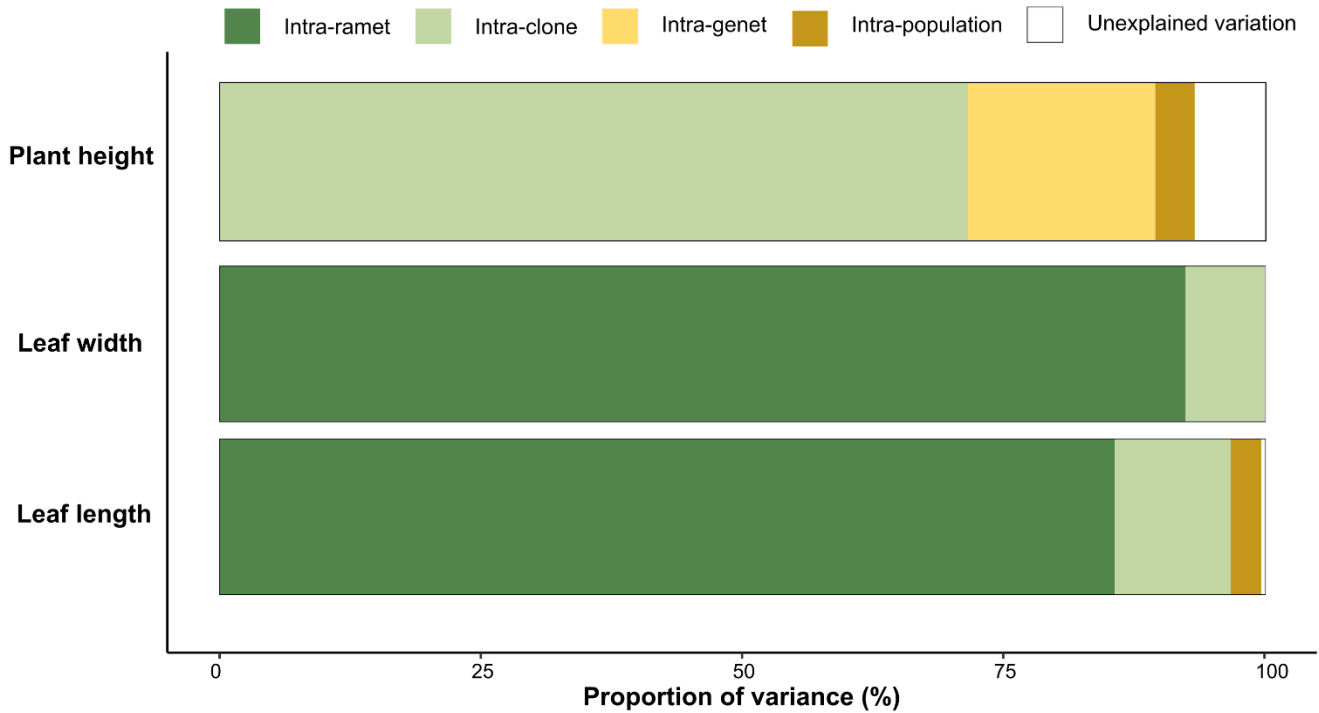
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40 **Fig. 1.** Schematic figure of the hierarchical levels of organisation of *Galium odoratum*. In each population,  
 41 five genets were sampled allowing to calculate intra-population coefficient of variation (CV). Each genet  
 42 was separated by cutting its stolons into four separate ramets, that were each planted in their own individual  
 43 pot, which allows to calculate intra-genet CV. Each pot was under one of the four treatments (control shade  
 44 + no drought, early shade + no drought, control shade + drought, early shade + drought). Over time, the  
 45 individually planted ramets in each pot reproduced vegetatively and became a clonal individual made up of  
 46 several ramets, allowing to calculate intra-clone and intra-ramet CV, respectively. The CV abbreviation is  
 47 shown at each hierarchical level.

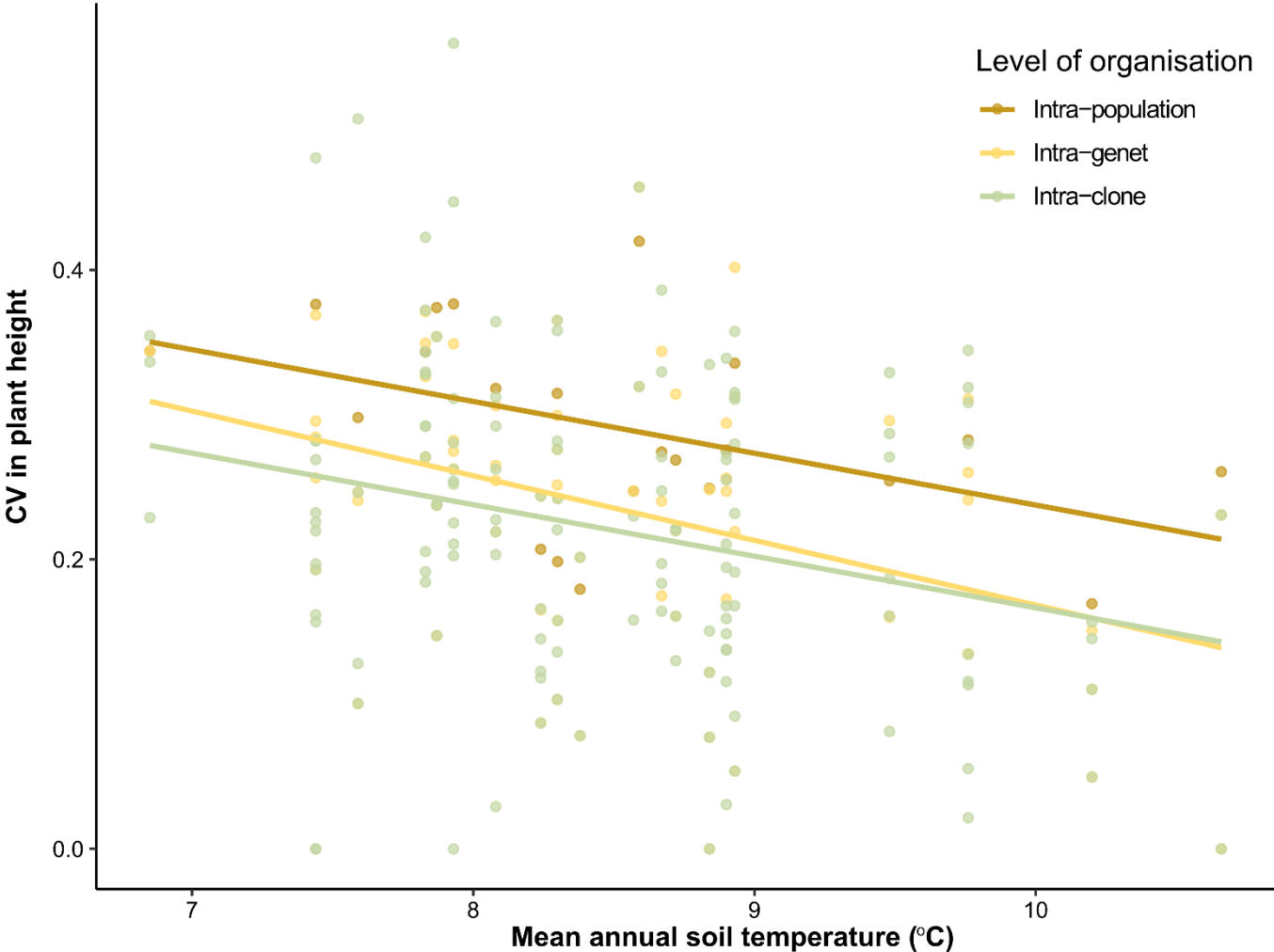
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50

51 **Fig. 2.** Proportion of variation in plant height, leaf length and leaf width of *Galium odoratum* explained by  
52 variance partitioning among different levels of organisation: intra-population, intra-genet, intra-clone, intra-  
53 ramet or unexplained.

54



57 **Fig. 3** Relationship between mean annual soil temperature at the site of origin with the intra-individual  
58 variation in height in *Galium odoratum*. Regression lines through data (dots) from different hierarchical  
59 levels of organisation are shown.



61

**A) Intra-clone**

Shade treatment	CV in leaf length (approx. mean)
Control shade	0.472
Early shade	0.506

**B) Intra-ramet**

Treatment	CV in leaf width (approx. mean)
Control shade + No drought	0.416
Early shade + No drought	0.404
Control shade + Drought	0.370
Early shade + Drought	0.420

62

63 **Fig. 4.** Shading and drought treatment effects on the coefficient of variation (CV) at two intra-individual  
 64 levels in *G. odoratum*. A) Effect of control and early shade treatment on CV in leaf length at the intra-clone  
 65 level. B) Effect of the four different treatment combinations on CV in leaf width at intra-ramet level.

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The figure is a scatter plot with two quadratic regression lines and shaded confidence intervals. The y-axis is labeled 'Mean leaf length (mm)' and ranges from 10 to 30. The x-axis is labeled 'CV in leaf length' and ranges from 0.2 to 0.7. A legend in the top right corner indicates 'Drought' status: 'Yes' (blue) and 'No' (orange). The 'Drought: Yes' data points are blue dots, and the 'Drought: No' data points are orange dots. The blue quadratic curve shows a downward trend, starting at approximately 19.5 mm at CV=0.2 and ending at approximately 12.5 mm at CV=0.7. The orange dashed quadratic curve is relatively flat, starting at approximately 17.5 mm at CV=0.2 and ending at approximately 16.5 mm at CV=0.7. Shaded areas around the curves represent 95% confidence intervals.

68

69 **Fig. 5.** Significant two-way interaction between the drought treatment and quadratic intra-clone CV in leaf  
70 length on the mean leaf length in *Galium odoratum*. The dashed and orange curve represents the control  
71 treatment while the blue curve represents the drought treatment. 95% confidence intervals are shown around  
72 the curves.

34