Trait plasticity and covariance along a continuous soil moisture gradient J. Grey Monroe^{1,4}, Haoran Cai², & David L. Des Marais^{2,3}

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Water is perhaps the greatest environmental determinant of plant yield and fitness, yet our understanding of plant-water relations is limited because it is primarily informed by experiments that have treated soil moisture variability as two simple discrete levels - wet and dry - rather than a continually varying environmental gradient. Here we used experimental and statistical methods based on function-valued traits to study responses to continuously varying soil moisture gradient in a suite of physiological and morphological traits in two species and multiple genotypes of the model grass *Brachypodium*. We find that the majority of traits exhibit non-linear responses to soil moisture variability. We also observe differences in the shape of these non-linear responses between traits, species, and genotypes. Emergent phenomena arise from this variation including changes in trait correlations and evolutionary constraints as a function of soil moisture. These results point to the importance of considering non-linearity in plant-water relations to predict plastic and evolutionary responses to changing climates.

Keywords: drought, function-valued traits, Brachypodium, evolutionary constraint, phenotypic plasticity, non-linearity

Introduction

For plants, soil water availability is one of the most important environmental factors in ecology and agriculture, acting as a major determinant of fitness and yield (Juenger 2013; Greenham et al. 2017). Considerable interest and effort have been placed on studying plant responses to drought, often defined conceptually and experimentally as an environmental condition of abnormally elevated aridity resulting in decreased plant performance (Passioura 1996). Most of the vast research on drought responses and tolerance strategies, including in Brachypodium, the focal system of this work, has been limited to comparisons between discrete soil water levels - control and water-limited (Des Marais et al. 2012; Edwards et al. 2012; El-Soda et al. 2014; Vasseur et al. 2014; Greenham et al. 2017). Yet soil moisture as an environmental factor is complex and multidimensional, with fluctuations varying continuously in timing, duration, and degree. Here we investigate trait responses to one important dimension of soil moisture variability - degree - with experimental and statistical approaches treating soil moisture content as a continuous variable rather than a set of fixed levels.

The limited extent to which continuous environmental vari-

ables have been studied is not unique to investigations of plant responses to soil moisture. Recent decades have seen increased appreciation for modeling traits as mathematical functions of continuous variables across diverse organisms (Pettay et al. 2008; Rocha and Klaczko 2012; Mason et al. 2020), a wide range of traits (Robinson et al. 2009; McGuigan et al. 2010; Stinchcombe et al. 2010), and numerous components of the environment (Brommer et al. 2008; McGuigan 2009; Pearse et al. 2019). This burgeoning interest in the importance of "function-valued traits" has evolved from a call-to-arms to a field experiencing rapid theoretical and empirical advances (Kirkpatrick and Heckman 1989; Kingsolver et al. 2001; Griswold et al. 2008; Stinchcombe et al. 2012; Goolsby 2015; Gomulkiewicz et al. 2018). For plant biologists this presents a compelling framework for gaining a deeper understanding of plant adaptations to the environment.

A deep body of literature aims to understand the traits and strategies used by plants to acclimate to variable soil water content. Quantitative genetic variation in myriad drought resistance traits have been observed in natural populations and laboratory model systems. In particular, natural populations of Brassicaceae species (including *Arabidopsis thaliana*) and *Brachypodium* harbor variation both constitutive and inducible traits mediating plant-water relations, including water use efficiency (Des Marais *et al.* 2012, 2017; Edwards *et al.* 2012; Greenham *et al.* 2017), leaf chemistry (Kesari *et al.* 2012; Des Marais *et al.* 2017), leaf anatomy (Skirycz *et al.* 2011; Verelst *et al.* 2013; Dittberner *et al.* 2018), rootshoot biomass partitioning (Des Marais *et al.* 2012, 2017),

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and many others (Verslues and Juenger 2011; Edwards et al. 2012; Juenger 2013; Luo et al. 2016; Yarkhunova et al. 2016; Lenk et al. 2019). Among the numerous plant traits that can and have been measured, few have been studied more extensively than specific leaf area (SLA - often reported as it's inverse, leaf mass per area or LMA). SLA provides a description of leaf architecture that is central to the leaf economics spectrum, a theory which seeks to explain variation in leaf physiological strategies, from more conservative (low SLA) to more productive (high SLA) (Wright et al. 2004). In the context of drought stress, it is intuitive that low SLA might be adaptive as lower leaf surface area is expected to reduce water loss through transpiration. In several cases, reductions in SLA have been reported under drought conditions (Casper et al. 2001). For all of these traits, the shape of plastic responses to variation in water availability remain largely unmeasured.

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Despite abundant evidence of natural variation in organismenvironment interaction, little is known as to how trait covariance changes across continuous environmental gradients (but see Robinson et al. 2009). Describing this structure is important as trait variances and covariances can result in differential evolutionary constraints in alternative environments. For example, if total genetic variance in trait space changes depending on the environment, then the capacity to respond to selection will vary accordingly, with reduced responses to selection under conditions where trait variation is lower and vice versa. Similarly, if trait covariances depend on environment, conditions which increase trait covariances may limit evolutionary potential across a range of environments by reducing the effective axes of variation (Levins 1968; Via and Lande 1985; Kingsolver and Gomulkiewicz 2003; Gomulkiewicz et al. 2018). These phenomena are made more complex by the possibility that the relationship between trait variances and covariances with the environment may be nonlinear. Investigating complexities of genetic architecture is therefore needed for developing more complete quantitative predictions of evolvability in rapidly changing environments. Fortunately, advances have been made in statistical methods to reduce the dimensionality of genetic variance covariance matrices and produce meaningful summaries describing evolutionary constraints (Houle 1992; Blows 2007; Kirkpatrick 2009; Kingsolver et al. 2015).

In this study we combined these approaches with functionvalued trait analyses to investigate trait plasticity, covariance and evolutionary constraints across soil moisture gradients. We model trait responses to a continuous gradient of soil moisture for a range of genotypes of two species of the model grass genus *Brachypodium*. Because the shapes of trait responses cannot necessarily be known *a priori*, we use model selection among linear and non-linear environmental predictors to estimate the response function for each trait. We then estimate genotype means at different levels of soil moisture and compute the variance-covariance parameters for all traits. Finally, we ask whether patterns of variance and covariance in drying-responsive traits in *Brachypodium* species may lead to variation in evolutionary constraint as a function of soil moisture.

Materials and Methods

Genotypes and species. *Brachypodium* is a model genus for the genetics and genomics of C3 grasses (Brkljacic *et al.* 2011). In this work we studied natural variation between and among two species of *Brachypodium*: the annual *B. distachyon* and the perennial *B. sylvaticum*. Both species are endemic to Eurasia, with *B.distachyon* more prevalent in seasonally dry habitats in Southern Europe, North Africa, and the Middle East, and *B. sylvaticum* more widely distributed throughout Eurasia (Catalan *et al.* 2016) (Figure S1).

Five genotypes of each species were studied to characterize patterns of variation in plant traits across an environmental gradient: *Brachypodium distachyon* inbred lines ABR2, Adi-10, Bd21, Bd3-1, and Koz-1 and *Brachypodium sylvaticum* inbred lines Ain-1, Ast-1, Kry-1, Osl-1, Vel-1. For each species, these genotypes represent a range of geographical origins and phenotypic diversity (Steinwand *et al.* 2013; Des Marais *et al.* 2017). Both species are self-compatible and each of the lines used here have been maintained as inbred lines for greater than six generations (Vogel *et al.* 2009; Steinwand *et al.* 2013); as such, experimental replicates may be considered nearly homozygous.

Plant growth and dry down experiment. Plant growth and experimental soil dry down were performed in the greenhouses of the Arnold Arboretum of Harvard University. To synchronize germination across genotypes within each species, seeds were placed on damp filter paper in the dark at 4°C for 14 days prior to planting. To synchronize the developmental stage at the timing of the drought treatments between the two species, B. sylvaticum seeds were planted thirteen days before B. distachyon (Oct 7 and 20, 2015, respectively). For each genotype, 1200 seeds were planted two to a pot and were subsequently thinned to one plant, for a total of 600 experimental plants in a randomized block design. All plants germinated within four days of sowing. Individual seeds of plants were sown in Greens Grade Profile porous ceramic rooting media (Profile Products, Buffalo Grove, IL, USA) in Deepot D40H Conetainers (650mL; Stuewe & Sons. Tangent, OR, USA) and grown at 25°C/20°C days/nights. Ambient sunlight was supplemented to 1000 umol/m2/s for 12hr/day.

Dry down treatments began 29 and 42 days after sowing (DAS) for *B. distachyon* and *B. sylvaticum*, respectively. Because the harvesting was divided over five consecutive days (see section below), plants were split into five equal harvest

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Figure 1. Effect of the experimental dry down on soil water content. (A) Time series of gravimetric soil moisture for all pots during the 14-day dry down period. (B) Distribution of final (day 14) soil moisture content across all pots. The data are distinguished by color according to the watering treatment. (C) Final soil moisture content by genotype.

cohorts, with each cohort containing equal numbers of each watering treatment to avoid confounding harvest day with soil moisture content. Thus, though each consecutive cohort differed in age by a single day, each experienced the dry down treatment for the same amount of time. Nevertheless, we expected the difference in age between harvest cohorts to potentially impact trait expression and we therefore included harvest day (cohort) as a covariate in subsequent models. To generate a continuous gradient of final soil moisture by the end of the dry down period, plants were split into five watering treatments, receiving approximately 0, 4, 8, 12, 16, or 20 ml of water per day for 14 days. Prior to initiating the experiment pots were weighed with dry soil (mass_{dry}) and field capacity $(mass_{max})$ at the beginning of the experiment. These measurements provide the basis for calculating the gravimetric soil moisture content on each day and at the conclusion of the experiment. During the course of the dry down experiment, soil moisture content was calculated during the morning of day d for each pot as $mass_d/(mass_{max} - mass_{dry})$.

Plant harvesting and phenotyping. To characterize phenotypic responses to our experimental soil moisture gradient, we measured a suite of developmental and physiological traits. Plants were harvested in five cohorts over five days. Each day, half of the sampled plants were harvested for above and below ground biomass, total above ground green area, δC_{13} , N content, C content. The other half were harvested and assessed for specific leaf area (SLA) and relative water content (RWC). Above ground leaf area was estimated by laying freshly harvested plants flat between plates of clear plexiglass and imaging with a Nikon 5300 digital camera at a fixed distance with a 35mm Nikkor lens. Total green pixels were counted for each image with Easy Leaf Area (https://github.com/heaslon/Easy-Leaf-Area) with settings shown in Figure S3. Above ground biomass was measured after drying leaf material overnight at 60°C and then for several weeks at room temperature. Below ground

biomass was measured after washing the soil matrix from roots and drying them overnight at 60°C and then for several weeks at room temperature. Above and belowground biomass was measured after leaves and roots were dried. Leaf tissues for δC_{13} , δN_{15} , nitrogen (hereafter "N") content, and carbon (hereafter "C") content were ground to a fine powder and processed by the UC Davis Stable Isotope Facility. SLA was calculated by scanning the two youngest fully emerged leaves with a 1 cm^2 red square. Leaf area in mm^2 was calculated from these same images using Easy Leaf Area. SLA was calculated as *leaf area/biomass_{dry}*. These leaves were also used to calculate RWC. Prior to drying, fresh leaves were weighed (biomass fresh) and then submerged under water in 15mL falcon tubes for several hours. They were then weighed (biomassturgid), oven-dried overnight, and weighed again (biomass_{dry}). RWC was calculated as (biomass_{fresh} – biomass_{dry})/(biomass_{turgid} – biomass_{dry}).

Analyses

We used R for all statistical analyses. Code and data to generate this manuscript can be found at https://github.com/greymonroe/brachypodium_fvt.

Function-valued traits. For the purposes of modeling phenotypic responses to variation in soil moisture content, we considered soil moisture content as the final soil moisture on day 14 of the dry down period for each plant, referred to in figures as *Soil moisture*($\%_{final}$). A major challenge in studying function-valued traits is model selection. That is, identifying the functions that best describe the curvature (or lack thereof) in the shape of phenotypic responses to environmental gradients. Quadratic and natural splines have been suggested as potential functions to model non-linearities (Meyer 2005), but assuming the appropriate function is problematic. Akaike information criterion (AIC) selection based on contrasting multiple complex models offers an effective

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Table 1

Model selection. *** = predictor variable p < 1e-5, ** = predictor variable p < 1e-3, * = predictor variable p < 0.5, - = predictor variable included in selected model but p > 0.05. H = harvest day, S = species, G = genotype, E = final soil moisture, $E^2 =$ quadratic parameter, ns(E) = 2nd degree natural spline parameter

	G	Е	E^2	ns(E)	Н	G*E	G*E^2	G*ns(E)
RWC - B. distachyon			***	***	*			
RWC - B. sylvaticum	*			***	*			
SLA - B. distachyon	-		***	***	***		*	
SLA - B. sylvaticum	***			-				
Green Area - B. distachyon	**			***	***			
Green Area - B. sylvaticum	***		*		**		-	
Shoot Mass - B. distachyon	***		***		***			
Shoot Mass - B. sylvaticum	***		-		*			
Root Mass - B. distachyon	***		***	-	*			-
Root Mass - B. sylvaticum	***							
Root:Shoot - B. distachyon	***			*	***			
Root:Shoot - B. sylvaticum	***		*		***		*	
Biomass - B. distachyon	***			***	***			
Biomass - B. sylvaticum	***				*			
C content - B. distachyon			*					
C content - B. sylvaticum	*			-				
d13c - B. distachyon	***		***		***			
d13c - B. sylvaticum	***			**	***			
N content - B. distachyon	***		-		*			
N content - B. sylvaticum	***			***				**
d15n - B. distachyon	**			**	***			
d15n - B. sylvaticum			***	-	**			
C:N ratio - B. distachyon	***		**		*			
C:N ratio - B. sylvaticum	***		***				*	

means to balance predictability with over-fitting (Griswold *et al.* 2008; Gomulkiewicz *et al.* 2018). Thus, we began with the complex model for traits as described below.

 $Trait = H + G + E + E^2 + ns(E)_{df=2} + G * E + G * E^2 + G * ns(E)_{df=2}$

where H = harvest day, G = genotype, E = soil moisture, $E^2 = quadratic parameter$, $ns(E)_{df=2} = second degree natural spline parameter$

We then selected a model for each trait using stepwise AIC model selection with the *stepAIC* function from the package *MASS* (Venables and Ripley 2002) in R with the "direction" parameter set to "both." The two species were analyzed separately to avoid biases introduced by enforcing the same model on species with different sizes, developmental trajectories and evolutionary histories.

Genetic correlations as a function of soil moisture content. We calculated trait correlations at different levels of soil moisture to characterize how genetic correlations between traits vary as a function of soil moisture content. Predicted genotypic means for each trait were calculated at 20 levels of soil moisture content (from 0.3 to 1.0 gravimetric water content) based on the model chosen by AIC (see above). Next at each level of soil moisture, pairwise Pearson correlation coefficients between genotype means were calculated within each species.

Plasticity through multidimensional trait space. We quantified total plasticity through multidimensional trait space as a function of soil moisture by scaling each trait to a mean of 0 and calculating distance matrix between genotype means at all soil moisture levels. We looked specifically at total plasticity between consecutive soil moisture levels for each genotype. At each level of soil moisture, we then compared the two species by T-tests. To visualize plasticity of each genotype through multivariate trait space further, we performed a principal component analysis from the matrix of scaled genotype trait means using the *prcomp* function in R.

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Analysis of evolutionary constraints among traits. We calculated several statistics summarizing evolutionary constraint as described in Kirkpatrick (2009). First, for each species, we estimated the *G* matrix of genetic covariances between mean scaled genotype trait means at different levels of soil moisture. We then calculated, using the *prcomp* function in R, the eigenvalues of each mean standardized (trait values divided by mean) *G* matrix, λ_i . From these we then calculated the *number of effective dimensions*, n_D , equal to the sum of the eigenvalues divided by the largest eigenvalue:

$$n_D = \sum_{n=1}^n \lambda_i / \lambda_l$$

We also calculated the *maximum evolvability*, e_{max} , equal to the square root of the largest eigenvalue, λ_l (Houle 1992; Kirkpatrick 2009):

$$e_{max} = \sqrt{\lambda_l}$$

Finally, we calculated the *total genetic variance* (Kirkpatrick 2009), equal to the sum of the eigenvalues of *G*:

$$v_T = \sum_{n=1}^n \lambda_i$$

Results

The dry down experiment resulted in a continuous soil moisture gradient. Across the six watering treatments, combined with random variation in water capacity of pots (Figure S4), the dry down period resulted in a continuous environmental gradient of final soil moisture, but with a higher frequency of plants near the driest extreme of soil moisture variation (Figure 1). This gradient provides the basis for analyzing phenotypes in relation to soil moisture treated as a continuous gradient rather than limited set of discrete factors.

The observed reduction in leaf relative water content under the driest conditions in both *B. distachyon* and *B. sylvaticum* indicates that at this extreme, plants were physiologically stressed (Figure 2A). Mean leaf RWC for plants in the 10% tail of soil water content was 85.21% which is drier than that observed in the dry treatment of Des Marais et al. (Des Marais *et al.* 2017). Additional observations made during the experiment such as leaf rolling, another symptom of dehydration stress, was evident in plants at the lowest water treatment by the end of the dry down period.



Figure 2. Variation in phenotypic responses to soil moisture gradient as function-valued traits. *B. sylvaticum* genotypes are colored orange and *B. distachyon* blue. (a) RWC (b) SLA (c) Green Area (d) Shoot Mass (e) Root Mass (f) Root:Shoot (g) Biomass (h) C content (i) d13c (j) N content (k) d15n (l) C:N ratio

Non-linearity in trait responses to soil moisture is pervasive. We evaluated the degree to which traits show linear or non-linear shapes using an AIC model selection approach from a full model which included quadratic and natural spline parameters relating soil moisture content to plant phenotypes. We observed significant ($\alpha < 0.05$) non-linear components (quadratic, spline, or both) in the final models for all of the

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Figure 3. **Plasticity through multivariate trait space.** (a) Plasticity across all traits was calculated as distance between scaled phenotype for each genotype between different levels of soil moisture. Box plots indicate species median and 25th and 75th percentiles with whiskers extending to 1.5 times the interquartile range. "*" indicates significant differences between species (t-test, $\alpha < 0.05$). Principal component analysis of scaled phenotypic responses to soil moisture gradient among genotypes of both species. Upper panels show genotype means across soil mositure content. Percent values in axis titles indicate percent variance explained by that principal component. Lower panels show eigenvectors of each trait. (b) PC1 and PC2. (c) PC3 and PC4.

traits which included an environmental (water content) predictor (Table 1, Figure 2) except N content in *B. distachyon*, C content in *B. sylvaticum*, and shoot mass in *B. sylvaticum* where non-linear environmental predictors were included in the final models chosen by AIC but did not significantly explain trait variance (p > 0.05). In *B. distachyon*, all of the traits included at least one non-linear environmental predictor. In contrast, SLA, total biomass and shoot mass did not include environmental predcitor in selected models in *B. sylvaticum*. Interestingly, of all the traits which were predicted by environment in both *B. distachyon* and *B. sylvaticum*, the shape – whether a quadratic versus a spline function was included in the model – differed between the two species with the exception of carbon to nitrogen ratio.

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When considering plasticity across multidimensional trait space (Figure 3), it appears that most of the variation is attributable to responses to low soil moisture *B. distachyon* which was significantly more responsive to low soil moisture values (Figure 3A). In contrast, *B. sylvaticum* was more responsive to extreme wet conditions than *B. distachyon*. Across PC2 and PC3, we observed, particularly in *B. sylvaticum*, that phenotypes were similar between extreme dry and extreme wet soil moisture contents. This similarity may be explained

by the quadratic parameters of trait functions where the curvature of trait responses leads to similar phenotypes at both environmental extremes.

Nearly all traits show significant genetic variation. We also tested whether there was significant natural variation for the traits measured between genotypes in each species by looking at the parameters in the final model for each trait. Interestingly, in across species, most traits (20/24) included a significant genotype term in the final model, indicating significant differences between genotypes in magnitude of traits across all levels of soil moisture (Table 1, Figure 2). Though not formally tested, for other traits there are clear distinctions between the two species. For example, δC_{13} was considerably higher in B. sylvaticum (Figure 2). For SLA, while B. distachyon showed a strong response to soil moisture, especially under the driest conditions, SLA in B. sylvaticum was not responsive to soil moisture. In contrast, B. sylvaticum appears to show a more dramatic response in leaf composition estimated by N content and C:N ratio.

Several traits show interactions between genotype and non-linear responses to the environment. Significant interactions between genotype and environmental parameters in a final model indicate the presence of genetic variation

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Figure 4. **Trait correlations as a function of soil moisture content.** Correlations were calculated among genotypes by species (blue = B. *distachyon*, orange = B. *sylvaticum*). Note that correlations are not shown for traits in species where genotype was not included in final model (Table 1).

for plasticity (GxE) (Via and Lande 1985). For those GxE interactions where the environmental parameter is non-linear, significant GxE indicates genetic variation for the shape of reaction norms. SLA showed a significant interaction between genotype and soil moisture in *B. distachyon*. In *B. sylvaticum*, nitrogen content, carbon to nitrogen ratio, and root to shoot ratio all showed significant interactions between genotype and

soil moisture (Table 1, Figure 2). In each of these cases, the interaction between genotype and environment involved a nonlinear environmental predictor, indicating not only variation for the magnitude of plasticity (i.e. slope) but also variation in the shape of responses.

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Figure 5. Contrasting patterns of evolutionary constraint between B. distachyon and B. sylvaticum. Summary statistics of evolutionary constraint as a function of soil moisture in B. distachyon (blue) and B. sylvaticum (orange). (a and b) The number of effective dimensions, n_D , estimates number of unconstrained axes of variation (c and d) The maximum evolvability, e_{max} , corresponds to the square root of the largest eigenvalue of the genetic covariance matrix. (e and f) The total genetic variance, v_T , is equal to the sum of the eigenvalues of the genetic covariance matrix.

Correlations between traits change as a function of soil moisture, often in a non-linear fashion. We examined correlations between genotype trait means across soil moisture for traits where genotypic differences were observed (i.e. genotype predictor in trait models). Certain traits were strongly correlated regardless of environment. For example, correlations near 1 were observed between biomass and green area in both species across soil moisture. More complex relationships between trait correlations and soil moisture are observed in other trait combinations. For traits with genotype by non-linear environment interactions (Table 1), trait correlations showed non-linear relationships with soil moisture as well. Because more of these interactions were found in *B. sylvaticum* the number of trait combinations showing non-linear

relations between correlations and soil moisture appears to be higher than in *B. distachyon* (Figure 4). In some cases, the relationship between soil moisture and trait correlation was dramatic. The correlation between C:N ratio and root:shoot ratio in *B. sylvaticum*, for example, varied from approximately 0.3 under the wettest environment to approximately -0.7 under the driest environment.

Evolutionary constraints differ as a function of soil moisture and show contrasting patterns between Brachy*podium* species. To assess evidence of evolutionary constraint on the sampled traits, we estimated and analyzed parameters of the genetic covariance matrix, G, in each species across the soil moisture gradient. These analyses revealed contrasting patterns of evolutionary constraint both in relation to soil moisture and between *B. distachyon* and *B. sylvaticum*. In *B. distachyon*, the number of effective dimensions (n_D, n_D) which estimates number of axes of variation unconstrained by covariance) was lower when soils were drier. In contrast, n_D was lower in B. sylvaticum when soils were wetter. The maximum evolvability (e_{max} , variance through largest eigenvector of multidimensional trait space) also showed opposite trends between the two species. Whereas in *B. distachyon* e_{max} was highest under the driest conditions, in B. sylvaticum e_{max} was highest under the wettest conditions. The same trend was seen in total genetic variance $(V_T, which summarizes all genetic$ variance through multidimensional trait space). These results indicate that B. distachyon has increased genetic variance under dry conditions as compared to wet conditions, but that natural selection may be more constrained to act on this variation due to covariance between traits. In contrast, our results suggest that B. sylvaticum has decreased genetic variance under dry conditions on which selection might act but that this variation is less constrained by covariance between traits.

Discussion

Environmental conditions vary along continuous gradients in space, time, and degree. Natural populations of plants may therefore be exposed to a range of values for any particular dimension of the environment. The efficacy of natural selection to shape evolutionary response of populations depends on the magnitude of genetic variation in response to these environmental gradients and on the genetic correlation between traits as a function of the environment. Here, we explicitly model trait variation in two plant species as a continuous function of soil water availability and consider how genetic variance and co-variance in these functional responses may affect the evolution of plant-environment interaction.

Non-linearity in soil moisture response is pervasive in *Brachypodium.* We found significant non-linearity in response to a soil moisture gradient for all measured traits in at least one of the two species sampled. The best-fit function

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for some traits was quadratic, while other traits showed more complex responses to the environment which were best fit by a spline function. These results offer new insights with respect to the study of plant response to soil drying. By focusing on the curvature of phenotypic response as the explicitly modeled trait, we avoid contrasts of trait values expressed at arbitrary levels of soil water content which may obscure different thresholds of response among the diverse genotypes under study. SLA in B. distachyon exhibits this pattern, as two accessions show a threshold-like response in decreasing SLA as soils become drier, and three accessions express their highest SLA at intermediate SWC. Leaf N content (on a leaf-mass basis) in B. sylvaticum likewise shows considerable diversity of response with two accessions expressing their lowest values at intermediate SWC, one accession expressing its highest Leaf N at intermediate SWC, and one accession showing a nearly linear response along the SWC gradient. Such nonlinearities in trait responses to soil moisture reinforce the need to consider the severity of extreme weather events when predicting plastic responses, especially when scaling up to investigating the ecosystem consequences of plant responses to environmental stress (Felton and Smith 2017).

Implications for evolution of Brachypodium. Leaf N and SLA are two axes of the classic Leaf Economic Spectrum (Wright et al. 2004) and so the contrasting responses of these traits between the annual B. distachyon and perennial B. sylvaticum may reflect broader differences in their life history strategies. We recently reviewed evidence for physiological, anatomical and developmental differences between herbaceous annual and perennial species, finding support for generally higher SLA in annuals, befitting a generally resourceacquisitive strategy (Lundgren and Des Marais). Garnier (Garnier 1992) argued that small changes in leaf anatomy (e.g. SLA) will likely have large effects on plant growth rate and resource use and could therefore tip the balance between perennial and annual strategies. We also found that signatures of evolutionary constraint differ along our imposed soil water content gradient. Specifically, we find evidence of highest evolvability in multitrait space (measured by e_{max} and v_T) in B. distachyon under the driest soils. In contrast, B. sylvaticum exhibited evidence of greater evolvability by the same measures under the highest soil water content studied here. We speculate that the pattern observed could be a reflection of the different life history strategies of these two species. Annuality is considered a drought adaptive strategy characterized by escaping drought through phenology, by flowering before and remaining dormant as seeds during the most drought prone seasons (Friedman and Rubin 2015; Monroe et al. 2019) Thus, because of their life history, populations of annuals may actually experience fewer episodes of strong selection from extreme drought, which could explain why we find elevated genetic variance under these environments. In contrast, perennials such as B. sylvaticum are subjected to all seasons and

might therefore, experience more frequent episodes of selection caused by dehydration stress, despite paradoxically being in found in environments where droughts are less frequent on an annual basis. This pattern is consistent with the predictions of cryptic genetic variation revealed under environments where selection is less frequent or severe (Schlichting 2008).

Implications for breeding drought adaptation. We found that genetic correlations between traits can vary dramatically even over relatively small changes in soil moisture (Fig. 4). Responses to selection for drought tolerance may therefore depend on drought severity because of these patterns in genetic correlations. In the context of breeding, exploratory studies such as this may be valuable for identifying conditions for which genetic correlations are aligned with breeding objectives. Similarly, we found that signatures of evolutionary constraint varied across the environmental gradient, suggesting that responses to selection may be improved or restricted in accordance with patterns of constraint in relation to environment. Interestingly, we also found that some species may be more responsive to selection in a given environment based on patterns of constraint.

From a practical perspective this work highlights the value of function-valued trait approaches that may be extended to studying plant-water relations in agricultural settings. In this experiment, we investigated variation in plant responses to a gradient of soil moisture using six watering levels, which in combination with random variation in water capacity of experimental pots, produced a continuous gradient of soil moisture ranging from field capacity of the soil to strongly waterlimited. In the field, multiple watering regimes in combination with random variability between plots may produce similar gradients of soil moisture. Here, water content was measured gravimetrically. New sensing technologies may be useful for quantifying soil moisture in an analogous fashion, to define soil moisture quantitatively and then apply function-valued statistical approaches to contrast trait expression among genotypes. Finally, while in this experiment we used destructive phenotyping methods to measure traits, non-destructive (and high throughput) phenotyping will enable measurement of yield or fitness data as well, to examine explicit connections between trait variation and adaptation to different degrees of soil moisture.

Acknowledgements

We thank Chase Mason and Eric Goolsby for insightful conversations about function-valued trait approaches. This work was supported by an Eco-Evo-Devo Network training grant, USDA-NIFA Award 2014-38420-21801, and Max Planck Society support for JGM.

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Contributions

JGM and DD funded, planned, and conducted the experiment. JGM, HC, and DD contributed to analyses and writing.

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Supplement

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Table S1

	Df	Sum.Sq	Mean.Sq	F.value	PrF.	trait
(Day_14^2)	1.00000	1,749.17374	1,749.17374	113.13775	0.00000	Relative_WC b_dist
ns(Day_14,df=2)	2.00000	1,875.17736	937.58868	60.64387	0.00000	Relative_WC b_dist
Harv	4.00000	205.30059	51.32515	3.31975	0.01248	Relative_WC b_dist
Residuals	137.00000	2,118.09770	15.46057	NA 2 20507	NA 0.01128	Relative_WC b_dist
Jello $D_{2}(D_{2}) = 14 df - 2)1$	2.00000	525.29507	335 84253	3.39397	0.00000	Relative_wC b_sylv Relative_WC b_sylv
Is(Day_14,u1=2)1 Harv1	4 00000	268 15302	67 03826	2 81676	0.00000	Relative_WC b_sylv
Residuals1	121.00000	2.879.77449	23.79979	NA	NA	Relative WC b svlv
Geno1	4.00000	140.92870	35.23217	1.75884	0.14122	SLA b dist
(Day_14^2)1	1.00000	2,414.64758	2,414.64758	120.54272	0.00000	SLA b_dist
is(Day_14,df=2)2	2.00000	2,327.88690	1,163.94345	58.10575	0.00000	SLA b_dist
Harv2	4.00000	651.79297	162.94824	8.13461	0.00001	SLA b_dist
GenoI(Day_14^2)	4.00000	337.42496	84.35624	4.21119	0.00310	SLA b_dist
Residuals2	127.00000	2,543.99629	20.03147	NA	NA	SLA b_dist
ieno2	4.00000	1,572.00031	393.00008	12.05355	0.00000	SLA b_sylv
s(Day_14,df=2)5	2.00000	2 077 75004	81.09524	2.48/24 NA	0.08735 NA	SLA b_sylv
lesiduais5	122.00000	22 276 228 05160	52.00452 8.004.084.72700	5 44806	0.00043	showaround graanaraa h d
s(Day 14.df=2)4	2.00000	192.517.942.01652	96.258.971.00826	64.79115	0.00000	aboveground_greenarea b_d
larv3	4.00000	59.381.207.50609	14.845.301.87652	9.99225	0.00000	aboveground_greenarea b_d
esiduals4	134.00000	199.081.240.28441	1.485.680.89764	NA	NA	aboveground greenarea b d
Jeno4	4.00000	219,846,397.86896	54,961,599.46724	15.60576	0.00000	aboveground_greenarea b_s
Day_14^2)2	1.00000	24,060,407.68683	24,060,407.68683	6.83170	0.01005	aboveground_greenarea b_s
Iarv4	4.00000	77,050,113.87350	19,262,528.46838	5.46939	0.00043	aboveground_greenarea b_s
GenoI(Day_14^2)1	4.00000	33,700,623.60514	8,425,155.90129	2.39223	0.05412	aboveground_greenarea b_s
tesiduals5	126.00000	443,756,646.65128	3,521,878.14803	NA	NA	aboveground_greenarea b_s
ieno5	4.00000	151,815.09343	37,953.77336	17.16469	0.00000	Shoot_Mass b_dist
Day_14^2)3	1.00000	66,562.14458	66,562.14458	30.10291	0.00000	Shoot_Mass b_dist
larv5	4.00000	102,805.02689	25,701.25672	11.62346	0.00000	Shoot_Mass b_dist
lesidualso	136.00000	500,716.87291	2,211.15348	NA 24.41054	NA 0.00000	Snoot_Mass b_dist
enob	4.00000	282,513.50822	/0,628.37705	24.41054	0.00000	Snoot_Mass b_sylv
Day_14^2)4	1.00000	0,122.54739	0,122.54739	2.11607	0.14821	Shoot_Mass b_sylv
iaivo Pesiduale7	4.00000	370 349 46472	12,304.93393	4.54900 NA	0.00248 NA	Shoot Mass b_sylv
leno7	4 00000	570,549.40472	2,073.33319	23 75364	0.00000	Root Mass b diet
(Day 14^2)5	1.00000	13 857 56690	13 857 56690	24.99316	0.00000	Root Mass b_dist
s(Day_14_2)5 s(Day_14_df=2)5	2 00000	2 853 05835	1 426 52918	2 57285	0.08032	Root Mass b_dist
arv7	4.00000	6.739.85641	1,684,96410	3.03896	0.01976	Root Mass b dist
ienons(Day 14.df=2)	8.00000	8,723,71652	1.090.46457	1.96673	0.05586	Root Mass b dist
esiduals8	126.00000	69.861.25916	554,45444	NA	NA	Root Mass b dist
ieno8	4.00000	176,944.42509	44,236.10627	24.22780	0.00000	Root Mass b sylv
esiduals9	133.00000	242,836.80767	1,825.84066	NA	NA	Root_Mass b_sylv
Jeno9	4.00000	2.81680	0.70420	36.70369	0.00000	Shoot_Root_Ratio b_dist
s(Day_14,df=2)6	2.00000	0.15610	0.07805	4.06809	0.01925	Shoot_Root_Ratio b_dist
Iarv8	4.00000	1.14490	0.28623	14.91840	0.00000	Shoot_Root_Ratio b_dist
tesiduals10	135.00000	2.59012	0.01919	NA	NA	Shoot_Root_Ratio b_dist
Jeno10	4.00000	2.15066	0.53767	58.14315	0.00000	Shoot_Root_Ratio b_sylv
(Day_14^2)6	1.00000	0.04624	0.04624	5.00027	0.02713	Shoot_Root_Ratio b_sylv
larv9	4.00000	0.38527	0.09632	10.41571	0.00000	Shoot_Root_Ratio b_sylv
ienol(Day_14^2)2	4.00000	0.13/8/	0.03447	3.72728	0.00671	Shoot_Root_Ratio b_sylv
Cesiduais I I	124.00000	1.14000	0.00925	NA 18.62845	NA 0.00000	biomass b. dist
s(Day 14 df=2)7	2,00000	147 377 96199	73 688 08000	17 55455	0.00000	biomass b_dist
arv10	4 00000	145 877 82793	36 469 45698	8 68793	0.00000	biomass b_dist
tesiduals12	135.00000	566,691,30055	4.197.71334	NA	NA	biomass b dist
Jeno12	4.00000	862.484.81413	215.621.20353	24,16680	0.00000	biomass b svlv
larv11	4.00000	98,371.42927	24,592.85732	2.75636	0.03065	biomass b sylv
esiduals13	129.00000	1,150,964.71805	8,922.20712	NA	NA	biomass b_sylv
(Day_14^2)7	1.00000	9,393.36268	9,393.36268	4.44262	0.03680	c_content b_dist
esiduals14	143.00000	302,355.81423	2,114.37632	NA	NA	c_content b_dist
ieno13	4.00000	16,887.94249	4,221.98562	4.20111	0.00314	c_content b_sylv
s(Day_14,df=2)8	2.00000	3,997.44433	1,998.72217	1.98884	0.14106	c_content b_sylv
esiduals15	128.00000	128,635.92473	1,004.96816	NA	NA	c_content b_sylv
eno14	4.00000	14.72235	3.68059	24.76573	0.00000	d13c b_dist
Day_14^2)8	1.00000	4.81710	4.81710	32.41303	0.00000	d13c b_dist
arv12	4.00000	/.60/00	1.90175	12.79639 NA	0.00000 NA	d13c b_dist
esiullais10	135.00000	20.00317	0.14802	INA 14.00060	INA 0.00000	d13c D_dist
s(Day 14 df=2)0	2 00000	3 81605	1 90802	14.09900 8.35607	0.00000	d13c b svlv
a(124)_14,01=2/9	2.00000	25 16097	6 29024	27 54772	0.00039	d13c b svlv
esiduals17	124.00000	28.31414	0.22834	NA	NA	d13c b svlv
Jeno16	4.00000	332.26225	83.06556	8.93282	0.00000	n content b dist
(Day 14^2)9	1.00000	15.55323	15.55323	1.67259	0.19812	n content b dist
Iarv14	4.00000	139.69914	34.92478	3.75579	0.00626	n_content b_dist
esiduals18	135.00000	1,255.35355	9.29892	NA	NA	n_content b_dist
ieno17	4.00000	456.11421	114.02855	25.06977	0.00000	n_content b_sylv
s(Day_14,df=2)10	2.00000	325.38436	162.69218	35.76872	0.00000	n_content b_sylv
enons(Day_14,df=2)1	8.00000	145.98218	18.24777	4.01187	0.00030	n_content b_sylv
tesiduals19	120.00000	545.81385	4.54845	NA	NA	n_content b_sylv
ieno18	4.00000	14.25665	3.56416	5.34810	0.00050	d15n b_dist
s(Day_14,df=2)11	2.00000	14.26839	7.13419	10.70500	0.00005	d15n b_dist
larv15	4.00000	21.75070	5.43768	8.15934	0.00001	d15n b_dist
tesiduals20	134.00000	89.30239	0.66644	NA	NA	d15n b_dist
(Day_14^2)10	1.00000	17.12283	17.12283	32.60495	0.00000	d15n b_sylv
s(Day_14,df=2)12	2.00000	2.16282	1.08141	2.05920	0.13179	d15n b_sylv
larv16	4.00000	13.73616	3.43404	6.53903	0.00008	d15n b_sylv
esiduals21	127.00000	66.69536	0.52516	NA	NA	d15n b_sylv
ieno19	4.00000	434.41850	108.60463	18.40460	0.00000	c_n b_dist
Day_14^2)11	1.00000	79.87162	79.87162	13.53538	0.00034	c_n b_dist
larv17	4.00000	75.29789	18.82447	3.19007	0.01537	c_n b_dist
tesiduals22	135.00000	796.62842	5.90095	NA	NA	c_n b_dist
ieno20	4.00000	1,484.86836	371.21709	40.40652	0.00000	c_n b_sylv
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(Day_14^2)12	1.00000	488.17121	488.17121	53.13683	0.00000	C_II D_SVIV

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Figure S1. Distributions of (a) B. distachyon and (b) B. sylvaticum reported on GBIF as of 2019.18.08. Examples of (c) B. distachyon Carly Slawson (CC BY 4.0, https://www.inaturalist.org/photos/42532397) and (d) B. sylvaticum Grzegorz Grzejszczak (CC BY-NC 4.0, https://www.inaturalist.org/photos/36088991) (GBIF.org (26 February 2020) GBIF Occurrence Download https://doi.org/10.15468/dl.rau5v9).

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3	4	17	14	10		28	23	40	31	21		55	42	48	43	41		72	64	67	73	69		81	88	85	98	99
6	7	9	8	20		35	22	36	27	34		56	54	51	47	57		66	80	70	74	65		92	87	94	86	100
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105	114	116	108	120		133	127	139	125	138		145	149	159	150	157		171	169	162	175	173		181	192	190	199	195
106	117	112	119	109		132	122	129	121	123		148	143	160	151	141	Ι	161	168	165	178	179	Ι	184	186	187	188	191
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209	219	204	206	218		234	226	237	228	230		243	244	257	254	250	t	264	266	267	268	271	t	289	299	284	286	298
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313 312 315	304 309 302	16 310 305 308	318 306	320 319 301	-	338 322 332	337 339 324	17 331 323 327	335 321	328 336 329	-	353 345 346	347 354 357	18 358 356	350 348 359	342 360	Ī	366 361 372	368 362 367	19 364 363	375 369 380	370 365 377	-	382 391 388	393 383 396	20 395 399	392 394	389 397 398
313 312 315 316	304 309 302 314	16 310 305 308 311	318 306 303 307	320 319 301 317	-	338 322 332 326	337 339 324 340	17 331 323 327 330	335 321 333 334	328 336 329 325	-	353 345 346 351	347 354 357 343	18 358 356 352 344	350 348 359 341	342 360 349 355	-	366 361 372 373	368 362 367 376	19 364 363 378 371	375 369 380 379	370 365 377 374	-	382 391 388 381	393 383 396 390	20 395 399 385 387	392 394 400 386	389 397 398 384
313 312 315 316	304 309 302 314	16 310 305 308 311	318 306 303 307	320 319 301 317	-	338 322 332 326	337 339 324 340	17 331 323 327 330	335 321 333 334	328 336 329 325	-	353 345 346 351	347 354 357 343	18 358 356 352 344	350 348 359 341	342 360 349 355	-	366 361 372 373	368 362 367 376	19 364 363 378 371	375 369 380 379	370 365 377 374	-	382 391 388 381	393 383 396 390	20 395 399 385 387	392 394 400 386	389 397 398 384
313 312 315 316	304 309 302 314	16 310 305 308 311	318 306 303 307	320 319 301 317	-	338 322 332 326	337 339 324 340	17 331 323 327 330	335 321 333 334	328 336 329 325	-	353 345 346 351	347 354 357 343	18 358 356 352 344	350 348 359 341	342 360 349 355	-	366 361 372 373	368 362 367 376	19 364 363 378 371	375 369 380 379	370 365 377 374	-	382 391 388 381	393 383 396 390	20 395 399 385 387	392 394 400 386	389 397 398 384
313 312 315 316	304 309 302 314	16 310 305 308 311 21	318 306 303 307	320 319 301 317	-	338 322 332 326	337 339 324 340	17 331 323 327 330 22	335 321 333 334	328 336 329 325	[- - -	353 345 346 351	347 354 357 343	18 358 356 352 344 23	350 348 359 341	342 360 349 355		366 361 372 373	368 362 367 376	19 364 363 378 371 24	375 369 380 379	370 365 377 374		382 391 388 381	393 383 396 390	20 395 399 385 387 25	392 394 400 386	389 397 398 384
313 312 315 316 413	304 309 302 314 418	16 310 305 308 311 21 417	318 306 303 307 414	320 319 301 317 416	-	338 322 332 326 440	337 339 324 340 428	17 331 323 327 330 22 433	335 321 333 334	328 336 329 325 436		353 345 346 351 456	347 354 357 343 460	18 358 356 352 344 23 453	350 348 359 341 444	342 360 349 355 442		366 361 372 373 471	368 362 367 376 463	19 364 363 378 371 24 465	375 369 380 379 464	370 365 377 374 466	- - - -	382 391 388 381 486	393 383 396 390	20 395 385 387 25 484	392 394 400 386	389 397 398 384 490
313 312 315 316 413 411	304 309 302 314 418 408	16 310 305 308 311 21 417 407	318 306 303 307 414 403	320 319 301 317 416 412	-	338 322 332 326 440 425	337 339 324 340 428 434	17 331 323 327 330 22 433 424	335 321 333 334 432 431	328 336 329 325 436 430	- - - - -	353 345 346 351 456 451	347 354 357 343 460 458	18 358 356 352 344 23 453 441	350 348 359 341 444 447	342 360 349 355 442 445		366 361 372 373 471 470	368 362 367 376 463 478	19 364 363 378 371 24 465 473	375 369 380 379 464 461	370 365 377 374 466 476		382 391 388 381 486 481	393 383 396 390 488 482	20 395 399 385 387 25 484 483	392 394 400 386 495 489	389 397 398 384 490 485
313 312 315 316 413 411 401	304 309 302 314 418 408 409	16 310 305 308 311 417 407 419	318 306 303 307 414 403 410	320 319 301 317 416 412 405	-	338 322 332 326 440 425 438	337 339 324 340 428 434 423	17 331 323 327 330 22 433 424 437	335 321 333 334 432 431 429	328 336 329 325 436 430 435	- - - - -	353 345 346 351 456 451 454	347 354 357 343 460 458 446	18 358 356 352 344 23 453 441 457	350 348 359 341 444 447 448	342 360 349 355 442 445 450		366 361 372 373 471 470 467	368 362 367 376 463 478 474	19 364 363 378 371 24 465 473 468	375 369 380 379 464 461 462	370 365 377 374 466 476 480		382 391 388 381 486 481 492	393 383 396 390 488 482 487	20 395 399 385 387 25 484 483 498	392 394 400 386 495 489 500	389 397 398 384 490 485 497
313 312 315 316 413 411 401 404	304 309 302 314 418 408 409 415	16 310 305 308 311 417 407 419 406	318 306 303 307 414 403 410 402	320 319 301 317 416 412 405 420	-	338 322 332 326 440 425 438 427	337 339 324 340 428 434 423 421	17 331 323 327 330 22 433 424 437 426	335 321 333 334 432 431 429 439	328 336 329 325 436 430 435 422	- - - - - -	353 345 346 351 456 451 454 459	347 354 357 343 460 458 446 452	18 358 356 352 344 23 453 441 457 449	350 348 359 341 444 447 448 455	342 360 349 355 442 445 450 443		366 361 372 373 471 470 467 475	368 362 367 376 463 478 474 472	19 364 363 378 371 371 465 473 468 479	375 369 380 379 464 461 462 477	370 365 377 374 466 476 480 469		382 391 388 381 486 481 492 493	393 383 396 390 488 482 482 487 496	20 395 389 385 387 25 484 483 498 498	392 394 400 386 495 489 500 499	389 397 398 384 490 485 497 494
313 312 315 316 413 411 401 404	304 309 302 314 418 408 409 415	16 310 305 308 311 417 407 419 406	318 306 303 307 414 403 410 402	320 319 301 317 416 412 405 420		338 322 332 326 440 425 438 427	337 339 324 340 428 434 423 421	17 331 323 327 330 22 433 424 437 426	335 321 333 334 432 431 429 439	328 336 329 325 436 430 435 422		353 345 346 351 456 451 454 459	347 354 357 343 460 458 446 452	18 358 356 352 344 23 453 441 457 449	350 348 359 341 444 447 448 455	342 360 349 355 442 445 450 443		366 361 372 373 471 470 467 475	368 362 367 376 463 478 474 472	19 364 363 378 371 24 465 473 468 479	375 369 380 379 464 461 462 477	370 365 377 374 466 476 480 469		382 391 388 381 486 481 492 493	393 383 396 390 488 482 482 487 496	20 395 385 387 25 484 483 498 491	392 394 400 386 495 489 500 499	389 397 398 384 490 485 497 494
313 312 315 316 413 411 401 404	304 309 302 314 418 408 409 415	16 310 305 308 311 417 407 419 406 26	318 306 303 307 414 403 410 402	320 319 301 317 416 412 405 420		338 322 332 326 440 425 438 427	337 339 324 340 428 434 423 421	17 331 323 327 330 22 433 424 433 424 437 426 27	335 321 333 334 432 431 429 439	328 336 329 325 436 430 435 422		353 345 346 351 456 451 454 459	347 354 357 343 460 458 446 452	18 358 356 352 344 453 441 457 449 28	350 348 359 341 444 447 448 455	342 360 349 355 442 445 450 443		366 361 372 373 471 470 467 475	368 362 367 376 463 478 474 472	19 364 363 378 371 24 465 473 468 479 29	375 369 380 379 464 461 462 477	370 365 377 374 466 476 480 469		382 391 388 381 486 481 492 493	393 383 396 390 488 482 487 496	20 395 399 385 387 25 484 483 498 491 30	392 394 400 386 495 489 500 499	389 397 398 384 490 485 497 494
313 312 315 316 413 411 401 404 502	304 309 302 314 418 408 409 415 513	16 310 305 308 311 417 407 419 406 26 515	318 306 303 307 414 403 410 402 512	320 319 301 317 416 412 405 420		338 322 332 326 440 425 438 427	337 339 324 340 428 434 423 421	17 331 323 327 330 22 433 424 437 426 27 526	335 321 333 334 432 431 429 439	328 336 329 325 436 430 435 422		353 345 346 351 456 451 454 459 553	347 354 357 343 460 458 446 452 558	18 358 356 352 344 23 453 441 457 449 28 557	350 348 359 341 444 447 448 455 554	342 360 349 355 442 445 445 443		366 361 372 373 471 470 467 475 580	368 362 367 376 463 478 474 472 568	19 364 363 378 371 465 473 468 479 29 573	375 369 380 379 464 461 462 477	370 365 377 374 466 476 480 469 576		382 391 388 381 486 481 492 493 591	393 383 396 390 488 482 487 496	20 395 385 387 25 484 483 498 491 30 585	392 394 400 386 495 489 500 499 584	389 397 398 384 490 485 497 494
313 312 315 316 413 411 401 404 502 511	304 309 302 314 418 408 409 415 513 503	16 310 305 308 311 21 417 407 419 406 515 519	318 306 303 307 414 403 410 402 512 514	320 319 301 317 416 412 405 420 509 517		338 322 332 326 440 425 438 427 535 533	337 339 324 340 428 434 423 421 530 527	17 331 323 327 330 433 424 433 424 437 426 526 539	335 321 333 334 432 431 429 439 537 525	328 336 329 325 436 430 435 422 528 538		353 345 346 351 456 451 454 459 553 551	347 354 357 343 460 458 446 452 558 558 548	18 358 356 352 344 453 441 457 449 28 557 547	350 348 359 341 444 447 448 455 554 554	342 360 349 355 442 445 445 445 443 556 552		366 361 372 373 471 470 467 475 580 565	368 362 367 376 463 478 474 472 568 574	19 364 363 378 371 24 465 473 468 479 29 573 564	375 369 380 379 464 461 462 477 572 571	370 365 377 374 466 476 480 469 576 570		382 391 388 381 486 481 492 493 591 590	393 383 396 390 488 482 482 487 496 583 598	20 395 385 387 25 484 483 498 491 30 585 593	392 394 400 386 495 489 500 499 584 581	389 397 398 384 490 485 497 494 586 596
313 312 315 316 413 411 401 404 502 511 508	304 309 302 314 418 408 409 415 513 503 516	16 310 305 308 311 417 407 419 406 515 519 505	318 306 303 307 414 403 410 402 512 514 520	320 319 301 317 416 412 405 420 509 517 518		338 322 332 326 440 425 438 427 535 533 533	337 339 324 340 428 434 423 421 530 527 522	17 331 323 327 330 22 433 424 433 424 437 426 526 539 529	335 321 333 334 432 431 429 439 537 525 521	328 336 329 325 436 430 435 422 528 528 528 523		353 345 346 351 456 451 454 459 553 551 541	347 354 357 343 460 458 446 452 558 548 549	18 358 356 352 344 453 441 457 449 28 557 547 559	350 348 359 341 444 447 448 455 554 554 554	342 360 349 355 442 445 450 443 450 443 556 552 545		366 361 372 373 471 470 467 475 580 565 578	368 362 367 376 463 478 474 472 568 574 563	19 364 363 378 371 24 465 473 468 479 29 573 564 577	375 369 380 379 464 461 462 477 572 571 569	370 365 377 374 466 476 480 469 576 576 577 575		382 391 388 381 486 481 492 493 591 590 587	393 383 396 390 488 482 487 496 598 598 594	20 395 385 387 25 484 483 498 491 585 593 588	392 394 400 386 495 489 500 499 584 581 582	389 397 398 384 490 485 497 494 586 596 600

Figure S2. Planting scheme.

Lea	af minim	um Gre	en RGB	value:	
	41				
0	50	100	150	200	250
	Leaf Gr 0.90	een Rat	io: (G/I	٤)	
	0.90	1.4	0	1.90	
	Leaf Gr	een Rat 1	io: (G/I 44	3)	
	0.80	1.30)	1.80	
Sca	ile minin 63	um Re	d RGB v	value:	
0	50	100	150	200	250
	Scale R	ed Ratio	: (R/G	& R/B) 1.9	5
	Scale R	ed Ratio	50 Store	& R/B) 1.9	5
Sca 1	Scale R 1.00 ale area (0	ed Ratio 1. cm^2):	50 (R/G	& R/B) 1.9 2.0	5
Sca 1	Scale R 1.00 ale area (0	ed Ratio 1. cm^2): 8.0	s: (R/G 50	& R/B) 1.9 2.0	5 00 20.0
Sca 1 0.0	Scale R 1.00 ale area (0) 4.0 Process	ed Ratio 1. cm^2): 8.0	s: (R/G 50 12.0	& R/B) 1.9 2.0 16.0	5)0 20.0
Sca 1 0.0	Scale R 1.00 ale area (.0) 4.0 Process	ed Ratio 1. cm^2): 8.0 ing Spe	50 12.0 ed:	& R/B) 1.9 2.0 16.0	5)0 20.0
Sca 1 0.0	Scale R 1.00 ale area (0 0 4.0 Process	ed Ratio 1. cm^2): 8.0 ing Spe	s: (R/G 50 12.0 ed:	& R/B) 1.9 2.0 16.0 4	20.0
Sca 1	Scale R 1.00 ale area (0) 4.0 Process	ed Ratio 1. cm^2): 8.0 ing Spe 2	s: (R/G 50 12.0 ed: 3	& R/B) 1.9 2.0 16.0 4	20.0
Sca 1 0.0	Scale R 1.00 ale area (0) 4.0 Process 1 nimum L	ed Ratio 1. cm^2): 8.0 ing Spe 2 eaf Size	: (R/G 50 12.0 ed: 3 t (pixel:	& R/B) 1.94 2.0 16.0 4 ;):	20.0
Sca 1 0.0	Scale R 1.00 ale area (.0 Process 1 nimum L)	ed Ratio 1. cm^2): 8.0 ing Spe 2 eaf Size	: (R/G 50 12.0 ed: 3 t (pixel:	& R/B) 1.94 2.0 16.0 4 (i):	5)0 20.0

Figure S3. Settings used in Easy Leaf Area.

FUNCTION-VALUED DROUGHT EXTREMES

17



Figure S4. Variation in pot field capacity.