

Alma Mater Studiorum Università di Bologna  
Archivio istituzionale della ricerca

Effect of climate extremes and grazing on functional traits of a grassland community: insights from a 20-year experiment

This is the final peer-reviewed author's accepted manuscript (postprint) of the following publication:

*Published Version:*

Pavanetto, N., Tordoni, E., Petruzzellis, F., Maccherini, S., Nardini, A., Beccari, E., et al. (2025). Effect of climate extremes and grazing on functional traits of a grassland community: insights from a 20-year experiment. *ANNALS OF BOTANY*, 136(1), 129-139 [10.1093/aob/mcaf059].

*Availability:*

This version is available at: <https://hdl.handle.net/11585/1029346> since: 2025-11-19

*Published:*

DOI: <http://doi.org/10.1093/aob/mcaf059>

*Terms of use:*

Some rights reserved. The terms and conditions for the reuse of this version of the manuscript are specified in the publishing policy. For all terms of use and more information see the publisher's website.

This item was downloaded from IRIS Università di Bologna (<https://cris.unibo.it/>).  
When citing, please refer to the published version.

(Article begins on next page)

1 **Annals of Botany**

2 Original Article

3 **Effect of climate extremes and grazing on functional traits of a grassland**  
4 **community: insights from a 20-year experiment**  
5

6 Nicola Pavanetto<sup>1§</sup>, Enrico Tordoni<sup>2§</sup>, Francesco Petruzzellis<sup>3,4§</sup>, Simona Maccherini<sup>5,6</sup>,  
7 Andrea Nardini<sup>3</sup>, Eleonora Beccari<sup>2</sup>, Carlo Ricotta<sup>7</sup>, Duccio Rocchini<sup>8,9</sup>, Luisa Conti<sup>9</sup>,  
8 Emanuele Fanfarillo<sup>5,6</sup>, Miris Castello<sup>3</sup>, Giovanni Bacaro<sup>3\*</sup>

9

10 <sup>1</sup>*Institute of Agricultural and Environmental Sciences, Estonian University of Life Sciences,*  
11 *51006 Tartu, Estonia*

12 <sup>2</sup>*Institute of Ecology and Earth Science, University of Tartu, J. Liivi 2, 50409 Tartu, Estonia*

13 <sup>3</sup>*Department of Life Sciences, University of Trieste, via L. Giorgieri 10, 34127 Trieste, Italy*

14 <sup>4</sup>*Department of Biology, University of Padova, via U. Bassi 58B, 35121 Padova, Italy*

15 <sup>5</sup>*Department of Life Sciences, University of Siena, via P.A. Mattioli 4, 53100 Siena, Italy*

16 <sup>6</sup>*NBFC, National Biodiversity Future Center, 90133 Palermo, Italy*

17 <sup>7</sup>*Department of Environmental Biology, University of Rome “La Sapienza”, Piazzale Aldo*  
18 *Moro 54, 00185 Rome, Italy.*

19 <sup>8</sup>*BIOME Lab, Department of Biological, Geological and Environmental Sciences, Alma*  
20 *Mater Studiorum University of Bologna, via Irnerio 42, 40126 Bologna, Italy*

21 <sup>9</sup>*Czech University of Life Sciences Prague, Faculty of Environmental Sciences, Department*  
22 *of Spatial Sciences, Kamýcka 129, 16500 Praha - Suchbátka, Czech Republic*  
23

24 **Running title**

25 Changes in grassland functional traits over 20-years experiment  
26

27 \* For correspondence. Email: [gbacaro@units.it](mailto:gbacaro@units.it)

28 § Nicola Pavanetto, Enrico Tordoni and Francesco Petruzzellis contributed equally to this  
29 work.  
30  
31  
32

1 **Abstract**

2 **Background and Aims:** Climate change, particularly the increased frequency of extreme  
3 climatic events, poses significant challenges to the biodiversity and functionality of semi-  
4 natural grasslands. However, the response of plant functional traits of grassland communities  
5 to climate extremes is still an unresolved issue. Using data from a long-term experiment, we  
6 aimed to characterize the functional response of a grassland community to simultaneous long-  
7 term effects of grazing and climate extremes.

8 **Methods:** For over a 20-year period, we monitored the species composition of grazed and  
9 ungrazed grassland plots. We measured functional traits defining the Leaf Economics  
10 Spectrum (LES) and the Hydraulic Safety-Efficiency (HSE) trade-offs, and we identified the  
11 temporal dynamics of single traits at the community level as well as the changes in functional  
12 strategies among grazed and ungrazed communities. Then, we assessed the effect of climatic  
13 extremes in driving the changes in functional composition.

14 **Key Results:** Grazed plots, in the first few years, were dominated by fast-growing species  
15 with more acquisitive strategies compared to ungrazed plots. However, both communities  
16 showed a reorganization in the functional structure over time, pointing towards a selection of  
17 trait combinations favoring more conservative, stress-tolerant strategies. The joint effect of  
18 grazing and climate extremes significantly altered the functional composition of the grazed  
19 community, leading to a shift from species with grazing-tolerant traits to species with  
20 grazing-avoidant, and drought-tolerant, traits.

21 **Conclusions:** We found that grazing pressure generally promoted functional diversity but led  
22 to rapid shifts in community composition when combined with prolonged drought events. In  
23 contrast, the ungrazed community, dominated by species with conservative resource-use  
24 strategies, showed more stable functional richness and divergence, as well as a reduced

1 sensitivity to climatic extremes. These results underscore the importance of carefully  
2 evaluating grazing in the context of climate change, particularly to guide restoration and  
3 conservation efforts.

4

5 **Keywords:** Calcareous grasslands, climate extreme events, community weighted means,  
6 drought, functional diversity, grazing, heatwaves, herbaceous plants, temporal dynamics,  
7 woody plants.

8

## 9 **Introduction**

10 Climate change has widespread and pervasive impacts on ecosystem integrity and  
11 biodiversity (Pörtner *et al.* 2022). Global warming, which is usually associated with increased  
12 intensity and frequency of extreme climatic events (Myhre *et al.* 2019), is negatively  
13 affecting ecosystem functioning particularly in the Mediterranean region, where the warming  
14 rate is expected to be about 20% larger than globally (Peñuelas *et al.* 2017; Lionello and  
15 Scarascia 2018; Petruzzellis *et al.* 2022).

16 European semi-natural grasslands, i.e. grasslands of anthropogenic origin represent  
17 one of the most species-rich ecosystems on the continent and an essential source of  
18 ecosystem services (Bengtsson *et al.* 2019; Dengler *et al.* 2020; Prangel *et al.* 2024). Despite  
19 their ecological importance semi-natural grasslands have experienced a dramatic decline in  
20 both diversity and spatial extent in recent decades (Bardgett *et al.* 2021). This decline is  
21 mostly attributed to overexploitation (Rota *et al.* 2017; Dengler *et al.* 2020) and to the  
22 abandonment of grazing (Bengtsson *et al.* 2019), as well as extreme climatic events such as  
23 drought (Harrison *et al.* 2018; Luo *et al.* 2019) and heatwaves (Qu *et al.* 2020).

1           Generally, the interaction between grazing and extreme climate events (*e.g.* prolonged  
2 drought events and heatwaves) may favor synanthropic and generalist species (Evans *et al.*  
3 2011; Smith 2011; Maccherini *et al.* 2018), which in turn is expected to alter both taxonomic  
4 and functional composition of grasslands (Carmona *et al.* 2012; Komac *et al.* 2015; Rota *et*  
5 *al.* 2017; Stampfli *et al.* 2018; Zhang *et al.* 2023; Lin *et al.* 2024). In particular, plant  
6 functional composition is assessed through measurements of plant functional traits, which are  
7 valuable proxies for determining plant adaptations to biotic and abiotic conditions (Westoby  
8 *et al.* 2002; Wright *et al.* 2004; Puglielli *et al.* 2022; Pavanetto, Niinemets, *et al.* 2024). Plant  
9 trait combinations can be summarized along major axes of variation (Díaz *et al.* 2016;  
10 Carmona *et al.* 2021; Puglielli *et al.* 2024), encapsulating strategies for resource allocation  
11 (Leaf Economic Spectrum - LES; Wright *et al.* 2004; Reich 2014) or water use (Hydraulic  
12 Safety-Efficiency - HSE, Petruzzellis *et al.* 2021). LES traits reflect the trade-offs among  
13 nutrient uptake, growth, and leaf lifespan (*i.e.* acquisitive vs conservative resource use  
14 strategies, Wright *et al.* 2004) and have been largely used to define grassland dynamics in  
15 response to climate variability (Sandel *et al.* 2010; Griffin-Nolan *et al.* 2019). LES traits  
16 received mixed support in relation to their capability to detect shifts in the functional  
17 strategies of grassland communities under prolonged drought or heatwaves (Hoover *et al.*  
18 2014; Elst *et al.* 2017; Luo *et al.* 2019; Vitra *et al.* 2019), suggesting that these traits alone  
19 might not be enough to assess grassland communities' responses to climate extremes (Luo *et*  
20 *al.* 2019). The inclusion of hydraulic traits, therefore, could be effective in capturing  
21 overlooked key mechanisms driving community changes (Anderegg *et al.* 2016).

22           Within the HSE framework, leaf water potential inducing cell turgor loss ( $\Psi_{tp}$ ) is  
23 considered a proxy of drought resistance, while according to the “flux trait network” (Sack *et*  
24 *al.* 2013), leaf venation traits (*i.e.*, vein length per unit leaf area, VLA) are mechanistically  
25 related to leaf hydraulic conductance ( $K_{leaf}$ ) and could be considered as proxies of leaf water

1 transport efficiency. The integration of these traits has elucidated patterns in woody plants  
2 and invasive species distributions and functions (Tordoni *et al.* 2019, 2020; Yan *et al.* 2020;  
3 Petruzzellis *et al.* 2021; Pavanetto, Carmona, *et al.* 2024), as well as grasslands' species  
4 response to drought (Sun *et al.* 2020; Kramp *et al.* 2022; Wang *et al.* 2024). Moreover, this  
5 integration has proven useful for understanding grasslands response to the combined effects  
6 of grazing and climate extreme events. Several studies found that  $\Psi_{tp}$  shifted toward higher  
7 values (i.e., lower drought resistance) in plots under drought stress (Griffin-Nolan *et al.*  
8 2019; Kramp *et al.* 2022), suggesting that drought avoidant species display higher survival  
9 rates under drought in grassland ecosystems (Sun *et al.* 2020). However, we still lack an  
10 evaluation of the combined effect of climate extremes and grazing since most of the previous  
11 studies consisted of manipulative experiments under controlled conditions run for a relatively  
12 short period (4 years on average), not accounting for the natural variability of environmental  
13 conditions, which strongly affects plant functional composition (Vitra *et al.* 2019; Griffin-  
14 Nolan *et al.* 2021).

15 In this study, we assessed the changes in functional composition and diversity of a  
16 semi-natural grassland subjected to grazing pressure over a 20-year period. Over this time  
17 interval, we also assessed the impact of extreme climatic events such as prolonged droughts  
18 and heatwaves and occurrences of severe precipitations and frost days. Using a suite of traits  
19 derived from LES and HSE frameworks, we aimed at i) highlighting the variation of  
20 functional traits distribution over time between a grazed and an ungrazed plant community;  
21 ii) evaluating the changes in functional structure (i.e., the patterns of organization of species  
22 within the functional space, Mouillot *et al.* 2011) of the two communities, using a  
23 multivariate approach to highlight temporal shifts in the functional space; iii) identifying  
24 which climatic extremes had the highest impact on functional traits change rate. We expected  
25 the functional diversity of the communities to change over time with different extents

1 between the grazed and ungrazed communities. We posited that climate extremes associated  
2 with water scarcity, such as droughts and heatwaves, would exert the strongest effect on  
3 driving changes in functional composition. Specifically, we expected that during years of  
4 reduced precipitation, species with a more conservative and drought tolerant strategy could  
5 better cope with reduced water availability, compared to species with a more acquisitive and  
6 drought avoidant strategy.

## 7 **Materials and Methods**

### 8 *Study area and vegetation sampling*

9 The experimental area, located near Monte Labbro (42°49'13" N, 11°31'33" E, WGS84) in  
10 Tuscany (Central Italy), lies in a semi-natural calcareous grassland that was subjected to  
11 restoration in the early spring of 2001 (see Detailed method S1 [**Supplementary**  
12 **Information**] for details). The experimental plots were located in a cleared overgrown  
13 pasture (before cutting, *Prunus spinosa* covered 80%), grazed by donkeys (0.5 ha<sup>-1</sup> year<sup>-1</sup>),  
14 which were reintroduced into the area few years before the restoration; the site is occasionally  
15 grazed by sheep, cattle and hares. In 2001, sixteen plots were established with a randomized  
16 block design, with four blocks and four 3×5 m experimental plots in each block. Each block  
17 contained plots randomly assigned to one of four treatments: (1) no grazing or sowing, (2)  
18 sowing without grazing, (3) grazing without sowing, and (4) combined sowing and grazing.  
19 Ungrazed plots were fenced in spring 2002. Previous studies on the same plots indicated a  
20 negligible effect of sowing compared to grazing (Maccherini and Santi 2012). Consequently,  
21 this study focuses solely on the grazing factor. Vegetation surveys started in 2002 and were  
22 conducted annually in the second half of June, within 1 × 2 m subplots centered in each  
23 experimental plot. No vegetation surveys were done in 2005, 2017 and 2018. All vascular  
24 plants were identified during data collection, and their abundance was expressed as ground  
25 cover using a point quadrat method with a density of 100 pins m<sup>-2</sup>. If shrubs were present in

1 the plots, when their height exceeded 1 m, their percentage cover was visually estimated on a  
2 scale ranging from 0 to 100%. Additionally, each plot was visually inspected for species not  
3 detected by pins, recording them with a nominal ground cover of 0.05%.

#### 4 *Functional trait measurements*

5 In late June 2019, we collected 146 samples to measure the functional traits of the  
6 species contributing up to 70% of the cumulative ground cover of each year from 2002 to  
7 2019 resulting in 31 sampled species. We decided to sample the most abundant (i.e.  
8 dominant, *sensu* Grime 1998) species in order to account for the species that are supposed to  
9 contribute the most to community and ecosystem processes (Sasaki and Lauenroth 2011; ;  
10 Pérez-Harguindeguy et al. 2013; Tordoni *et al.* 2019 but see also Avolio *et al.* 2019). We  
11 collected two to three whole individuals for each species and treatment (i.e., grazed or  
12 ungrazed). Following established methodologies (Pérez-Harguindeguy *et al.* 2013, but see  
13 also Petruzzellis *et al.* 2019, 2021), we measured a set of plant functional traits usually  
14 associated with the LES, HSE, leaf vein architecture and water-use efficiency. Specifically,  
15 we measured the following functional traits: specific leaf area (SLA, mm<sup>2</sup> mg<sup>-1</sup>); leaf dry  
16 matter content (LDMC, mg g<sup>-1</sup>); minor vein length per unit area (VLA, mm mm<sup>-2</sup>); water  
17 potential at turgor loss point ( $\Psi_{tp}$ , MPa); carbon isotope composition ( $\delta^{13}C$ , ‰), leaf  
18 nitrogen content (LNC, ‰), and leaf carbon content (LCC, ‰) A detailed procedure of  
19 functional trait measurements is reported in Supplementary Methods S2 [**Supplementary**  
20 **Information**]. The main summary statistics for each trait for each species and each species'  
21 relative abundance for each year are reported in supplementary Table S1-Table S2  
22 [**Supplementary Information**].

1 In addition, to qualitatively assess changes in plant community composition and  
2 function, we categorized the sampled species into four functional groups ('grass', 'forbs',  
3 'legumes' – including only the graminoids-competitor and nitrogen-fixer *Trifolium*  
4 *incarnatum* –, and 'shrubs') following established methodologies (Lavorel *et al.* 2007; Liira  
5 *et al.* 2008). Details on this classification are provided in detailed methods S3  
6 **[Supplementary Information]**, along with the relative ground cover values for the  
7 functional groups over the years (supplementary Figure S1 **[Supplementary Information]**).

### 9 *Climatic extremes*

10 We obtained the climatic data from the nearest meteorological station (S. Fiora, Centro  
11 Funzionale Regionale of Tuscany, 687 m a.s.l., 5.56 km away from the study area).  
12 Following the guidelines of the Expert Team on Climate Change Detection and Indices  
13 (ETCCDI) (Zhang *et al.* 2011), we computed temperature and precipitation indices from  
14 2002 to 2022. Specifically, we calculated the number of days with a minimum temperature  
15 ( $T_{\min} \leq 0^{\circ}\text{C}$ ) (FROST.DAYS) and the number of days with precipitation  $\geq 20$  mm  
16 (HEAVY.RAINS). Both indices were calculated for the period from the preceding December  
17 to June of each sampling year, aligning with the pre-growing season, which is crucial to  
18 assess the response of vegetation to such climatic drivers (Dukes *et al.* 2005). Additionally, to  
19 assess the impact of dry spells and heatwaves, we calculated the maximum number of  
20 consecutive days with precipitation  $< 1$  mm from January to June (DRY.DAYS) and the  
21 maximum number of consecutive days with temperatures  $> 30^{\circ}\text{C}$ , calculated from July of the  
22 previous year to June of the current year (HEATWAVES). This period captures the full scope  
23 of seasonal temperature and precipitation fluctuations relevant to plant growth and survival  
24 (Teuling *et al.* 2010). A detailed representation of these climatic trends over the study period  
25 is provided in supplementary Figure S2 **[Supplementary Information]**.

## 1 *Statistical Analysis*

### 2 *Changes in trait composition over time*

3 To evaluate changes in the functional composition, we calculated the community-weighted  
4 mean values (CWMs) of each functional trait for each year and plot using abundance-  
5 weighted kernel density distributions (Maitner *et al.* 2023). Specifically, we used a non-  
6 parametric bootstrapping method to include intra-specific variability through statistical  
7 resampling, following the workflow proposed by Maitner *et al.* (2023). To calculate the trait  
8 distributions and CWM values, we used the *trait\_np\_bootstrap* function in the R (v4.4.1, R  
9 Core Team, 2024) package ‘traitstrap’ (Maitner *et al.* 2023; Telford *et al.* 2023), which  
10 utilises random sampling with replacement in proportion to species abundance to generate  
11 new distributions reflecting the variation in observed data. We did this separately for the  
12 grazed and ungrazed community. To ease the interpretability and visualization of the  
13 temporal patterns, we grouped the trait distributions considering three temporal intervals  
14 (years 2002-2008, years 2009-2014, years 2015-2022); significant differences in the  
15 distribution of the values of each trait between the two communities were tested by  
16 performing a Kruskal-Wallis test followed by a post hoc Dunn’s test with Holm-Bonferroni  
17 correction for multiple comparisons, using the function *dunn\_test* in R package ‘rstatix’  
18 (Kassambara 2023). Probability density distributions of each trait across all years for both  
19 ungrazed and grazed communities are reported in Figures S3-S4 [**Supplementary**  
20 **Information**].

### 21 *Changes in the functional structure over time*

22 We evaluated the changes in the functional structure of the grazed and ungrazed community  
23 over the years using the Trait Probability Density framework (TPD - Carmona *et al.* 2016,  
24 2019, 2024). First, we calculated a trait space by performing a Principal Component Analysis

1 (PCA) considering each species' average functional trait values (supplementary Figure S5  
2 **[Supplementary Information]**). To evaluate the effect of phylogenetic relatedness in the  
3 position of the species within the trait space, we also performed a phylogenetically-corrected  
4 PCA (Revell 2009) which were later compared with the original PCA displaying a high  
5 degree of concordance (Procrustes correlation = 0.988,  $p = 0.001$ ; see Supplementary  
6 Methods S4 **[Supplementary Information]** for further methodological details. We then  
7 retained the first two principal components, explaining 62.81% of the total variance, to  
8 estimate the TPD at the community level (i.e., for each plot) for each year and treatment,  
9 weighting the trait probability of each species by their relative abundances (supplementary  
10 Figure S6-S7 **[Supplementary Information]**). For each of the three temporal intervals  
11 delineated in the previous section, we derived the functional richness (FRic, the area  
12 occupied by the trait space) and functional divergence (FDiv, the tendency of most abundant  
13 trait values to be closer to the margins of the functional space in case of high divergence and  
14 vice versa) at the 99% quantile threshold. To highlight the portions of the functional space  
15 that underwent major changes across years, we took advantage of the grid-based nature of the  
16 TPD framework to calculate cell-wise temporal variation expressed in terms of standard  
17 deviations, and we averaged them as explained in the previous section to obtain three  
18 temporal intervals. We finally estimated overlap-based dissimilarity among temporal  
19 intervals to assess the portions of the functional space occupied by the grazed and ungrazed  
20 community (i.e., dissimilarity) and quantify potential shared portions in the functional space  
21 (i.e., nestedness, Carmona *et al.* 2016) at the 99% and 50% quantile thresholds, using the  
22 same procedure as Beccari *et al.* (2024a).

23 To break the well-known relationship between the functional metrics and species  
24 richness (Beccari *et al.* 2024b), we also computed a null model as follows: i) we first  
25 transformed abundance-data in presence-absence data and we created 499 randomizations of

1 the community using the ‘Curveball algorithm’ (Strona *et al.* 2014), independently for grazed  
2 and ungrazed plots; ii) for each randomization we recalculated functional richness and  
3 divergence at the 99% of total probability; iii) we calculated standardized effect size as  $SES =$   
4  $\text{observed value} - \text{mean (randomizations)}/\text{standard deviation (randomizations)}$  and p-values by  
5 comparing the SES values with a cumulative normal distribution with mean = 0 and standard  
6 deviation = 1. For functional richness (sesFRic), negative SES values indicate that the species  
7 in each combination year-treatment tend to be more clustered in the functional space than  
8 expected by the same number of randomly selected species across the time-series in that  
9 treatment and vice versa. For functional divergence (sesFDiv), negative SES values would  
10 suggest a reduction in the prevalence of divergent/extreme values across the time series, with  
11 species converging toward the center of the functional space.

12 For temporal turnover, we used a slightly different approach to preserve the  
13 community structure across years while removing any directional temporal patterns, thus  
14 maintaining the ecological realism of the simulation process. Specifically, we used cycle-shift  
15 permutation, following the same approach described in Hallett *et al.* (2014, 2016), which  
16 preserves the temporal autocorrelation in the time series while breaking cross-correlations  
17 among species by reshuffling the starting time of each species along the time series in each  
18 permutation (Magurran *et al.* 2019). The same approach was used for dissimilarity metrics at  
19 99% and 50% quantiles, and we calculated SES as explained above. In this case, a significant  
20 SES implies that the functional structure is changing faster than expected by chance alone.

### 21 *Relationships between climatic extremes and changes in functional composition*

22 To assess the potential effect of the climatic extremes on functional composition, we first  
23 calculated the annual rate of change of CWMs (i.e., the trait velocity, Trugman *et al.* 2020)  
24 for each trait as the difference of CWM values between one sampling year and the successive

1 one, divided by the length of the sampling period. This approach allowed us to account for  
2 the gaps in the vegetation surveys and, since the climatic variables were calculated over the  
3 periods preceding each sampling year, permitted direct comparisons between the trait's  
4 annual rate of change and climatic extremes. After that, we fitted linear mixed models  
5 (LMMs), separately for the grazed and ungrazed community, setting trait velocities as the  
6 response variable and climatic extremes and plots as fixed and random effects, respectively.  
7 Plots were included as a random effect to control for the repeated measurement per plot  
8 (Martínez-Villa *et al.* 2024). Prior to the modelling procedure, we checked multicollinearity  
9 between climate predictors using Pearson's correlation coefficient (see supplementary Table  
10 S3 [**Supplementary Information**]). Each model retained all the climate extremes, as their  
11 correlations were lower than  $|0.7|$  (Dormann *et al.* 2013). Each predictor was standardized to  
12 zero mean and unit variance to obtain comparable coefficients. To account for second-degree  
13 non-linear relationship between the predictors and the response variables, we included a  
14 second-degree (quadratic) orthogonal polynomial term for each climate extreme. We checked  
15 whether to include or not the quadratic terms by performing multiple Likelihood Ratio Test  
16 (LRT) between models with and without quadratic terms by using the base R function  
17 'anova'. (p-value always  $< 0.001$  when including quadratic terms). All LMMs were  
18 performed using the REML method for the estimation.

## 19 **Results**

### 20 *Changes in trait composition over time*

21 The relative cover of the functional groups followed different temporal trends in grazed and  
22 ungrazed plots (supplementary Fig S1 [**Supplementary Information**]), which was reflected  
23 in the different community-weighted trait distributions. Indeed, trait distribution was  
24 significantly different between grazed and ungrazed communities across all temporal

1 intervals (Dunn's test always significant with  $p < 0.001$ ), except for LCC in the first two  
2 temporal intervals (**Fig. 1**). Grazed plots consistently showed higher values of SLA, VLA,  
3  $\Psi_{\text{tip}}$  and LNC compared to ungrazed plots (**Fig. 1b-d,g**). Conversely, LDMC and  $\delta^{13}\text{C}$  values  
4 were higher in ungrazed communities (**Fig. 1a, e**). Over the study period, both community  
5 types experienced a shift towards higher LDMC and lower SLA,  $\Psi_{\text{tip}}$ , and LNC, particularly  
6 marked in the final period (years 2015-2022). VLA values increased over time in the  
7 ungrazed community, while showing a slight decline in the grazed one. Both  $\delta^{13}\text{C}$  and LCC  
8 slightly increased in the last temporal interval in both grazed and ungrazed plots.

### 9 *Temporal variation of the functional structure*

10 The first axis (PC1) of the functional trait space (46.31% of total variance) was mainly  
11 related to SLA and LNC, and to LDMC and  $\delta^{13}\text{C}$ , representing a gradient between acquisitive  
12 and conservative resource-use strategies (**Fig. 2**, supplementary Table S4 [**Supplementary**  
13 **Information**]). The second axis (PC2), accounting for 16.5% of the total variance, was  
14 primarily associated with VLA,  $\Psi_{\text{tip}}$ , and LCC (**Fig. 2**), which further differentiated species  
15 based on drought tolerance and photosynthetic efficiency. Grazed plots consistently exhibited  
16 higher functional diversity (**Fig. 2 a,d,g**) than ungrazed ones (**Fig. 2 c,f,i**), with mean  
17 sesFRic of 8.05 (range 1.68 — 17.52) and 0.01 (-1.94 — 3.52), respectively (supplementary  
18 Table S5 [**Supplementary Information**]). Notably, we detected a steady increase in  
19 functional richness especially for ungrazed plots that shifted from negative to positive values  
20 of SES, suggesting an introduction of new functional strategies that expanded the functional  
21 diversity of these communities (supplementary Figure S6 [**Supplementary Information**]). In  
22 contrast, we observed a strong significant decoupling for sesFDiv between grazed (clustering)  
23 and ungrazed (dispersion) plots in the years 2015-2022, a pattern absent in earlier  
24 intervals (supplementary Table S6 [**Supplementary Information**]).

1 Over time, species in both grazed and ungrazed plots strongly reorganized in the  
2 functional space (**Fig. 2**). Both communities showed a consistent shift towards more  
3 conservative strategies, as highlighted by the patterns of temporal variation showing changes  
4 in the functional structure towards higher values of LDMC,  $\delta^{13}\text{C}$  and LCC, despite some  
5 nuances between treatments (**Fig. 2j,k**). Indeed, the temporal variability of ungrazed plots  
6 (**Fig. 2k**) is clustered in a smaller, central portion of the trait space suggesting minimal but  
7 consistent temporal changes promoting a higher stability over time, with limited shifts in trait  
8 composition. In contrast, grazed plots displayed higher standard deviations across the trait  
9 space, indicating substantial temporal fluctuations in functional composition (**Fig. 2j**).

10 Dissimilarities between the two treatments were significant from the seventh year for  
11 the whole functional space, whereas the high-density regions within the trait space (i.e., the  
12 ones in the 0.50 quantile) were more dissimilar than expected by chance in all comparisons  
13 (**[Supplementary Information]** Table S7). Moreover, negative values of nestedness indicate  
14 that trait combinations between treatments are minimally shared, further indicating a strong  
15 shift of the hotspots (i.e., high-density regions) within the trait space. This aspect can be  
16 visually appreciated by looking at the trait probability distributions across all years for both  
17 ungrazed and grazed communities in supplementary Fig. S6 – Fig. S7 **[Supplementary**  
18 **Information]**.

#### 19 *Relationships between Traits and Extreme Climate Variables*

20 The annual rate of change of the CWM of each trait was significantly associated with the  
21 climate variables (supplementary Table S8 **[Supplementary Information]**). In general, both  
22 communities showed a similar relationship between CWM's annual rates of change and  
23 climatic extremes. Specifically, SLA and LNC showed negative rates of change, while  
24 LDMC,  $\delta^{13}\text{C}$  and LCC exhibited positive rates of change in response to increasing dry spell

1 duration during spring (DRY.DAYS) in both grazed and ungrazed plots (**Fig. 3a-b,e-g**).  
2 Notably,  $\Psi_{tip}$  and VLA values remained stable with low to moderate DRY.DAYS but  
3 decreased during prolonged dry periods for both community types (**Fig. 3c-d**). Years with  
4 higher heatwaves duration (HEATWAVES) were generally associated with shifts in  
5 community trait towards higher values of SLA,  $\Psi_{tip}$ , VLA and LNC and lower values of  
6 LDMC and  $\delta^{13}C$  (**Fig. 3h-l, n**). Years associated with a higher number of heavy precipitation  
7 days (HEAVY.RAINS) showed little changes in community trait values for almost all trait  
8 considered; conversely, years with lower HEAVY.RAINS shifted towards lower values of  
9 LDMC and LCC and higher values of SLA,  $\Psi_{tip}$ , and LNC (supplementary Fig. S8  
10 **[Supplementary Information]**) in both community types. Model's coefficients for the  
11 grazed community were always higher than ungrazed plots, underscoring higher variability in  
12 trait response. Significant relationships for the number of frost days (FROST.DAYS) were  
13 found for a few traits only in the grazed community. However, they held negligible influence  
14 in traits' rates of change (supplementary Fig. S8 **[Supplementary Information]**).

## 15 **Discussion**

16 Grazing disturbances tended to favor species characterized by more acquisitive and fast-  
17 growing strategies, contrasting with the conservative, slow-growing strategies predominant  
18 among species in ungrazed plots. Over time, however, both the grazed and ungrazed  
19 communities experienced a significant reorganization in their functional structures,  
20 suggesting a combined influence of grazing and climate extremes in shaping their long-term  
21 structural and functional dynamics.

### 22 *Grazing and climate extremes drive changes in functional composition and diversity*

23 Perennial and annual herbs, including the annual nitrogen-fixer *Trifolium incarnatum*, were  
24 the dominant species in grazed plots. These species generally thrive only under grazing

1 conditions and compete with perennial grasses for land cover (supplementary Fig. S1  
2 **[Supplementary Information]**) (Hayes and Holl 2003). These species aligned with the  
3 acquisitive side of the LES (**Fig. 1a,b, Fig. 2**), minimizing leaf construction and maintenance  
4 costs while maximizing the ability to acquire resources and reproduce rapidly (Wright *et al.*  
5 2004; Reich 2014; Díaz *et al.* 2016). Furthermore, species in grazed plots tended to adopt  
6 strategies maximizing water transport efficiency at the expense of lower drought resistance  
7 compared to ungrazed plot (Nardini and Luglio 2014; Zhu *et al.* 2018). Consistently, grazed  
8 plots had lower  $\delta^{13}\text{C}$  and higher LNC CWM values (**Fig. 1e,g, Fig. 2**), indicating potential  
9 higher gas exchange and photosynthetic rates (Farquhar *et al.* 1989; Wright *et al.* 2004; Sack  
10 *et al.* 2013) than in ungrazed plots. This grazing-tolerance strategy reflects a competitive  
11 edge for species in grazed plots, characterized by rapid nutrient acquisition to compensate  
12 tissue loss through regrowth (Díaz *et al.* 2007; Kurze *et al.* 2021). Conversely, ungrazed  
13 plots were dominated by more conservative species such as *Bromus erectus*, which is a tall-  
14 growing grass well-adapted to drought that produces dense litter layers when abundant,  
15 potentially inhibiting other species' growth (Poniatowski *et al.* 2018). These species showed  
16 a more conservative growth strategy, developing durable leaves (higher LDMC, **Fig. 1a**) with  
17 higher drought stress resistance (lower  $\Psi_{\text{tip}}$ , **Fig. 1c**) at the cost of reduced water transport  
18 efficiency (lower VLA, **Fig. 1d**), photosynthetic capacity (lower LNC, **Fig. 1g**) and carbon  
19 fixation rates (higher  $\delta^{13}\text{C}$ , **Fig. 1e**).

20         However, the temporal variations of the functional structure highlighted that both  
21 grazed and ungrazed communities have undergone a consistent shift towards more  
22 conservative strategies, with a general increase in the diversity of strategy present (**Fig. 2**,  
23 supplementary Fig. S5 **[Supplementary Information]**). Specifically, the ungrazed  
24 community exhibited increased functional richness and functional diversity over time,  
25 suggesting an expansion in functional strategies. Species in the functional space were initially

1 clustered towards the center (supplementary Fig. S6 [**Supplementary Information**]), but  
2 over time they shifted into two different hotspots towards regions of the functional space  
3 associated with higher LDMC, VLA and LCC. This shift suggests the introduction of new  
4 specialized, more conservative strategies, thus leading to an increase in functional divergence  
5 and richness. This pattern was likely driven by the gradual encroachment of woody shrubs  
6 (e.g., *Acer monspessulanum*, *Prunus spinosa*, supplementary Fig. S1 [**Supplementary**  
7 **Information**]), whose trait suites and growth strategies significantly differ from those of  
8 herbaceous plants (Šímová *et al.* 2018), and occupy a different region in the global spectrum  
9 of plant form and function (Díaz *et al.* 2016). On the contrary we observed a gradual  
10 reduction in sesFDiv over time in grazed plots, where species positioning in the functional  
11 space has shifted from regions associated with grazing-tolerance strategies (higher SLA, LNC  
12 and  $\Psi_{tp}$ ) (Díaz *et al.* 2007) towards regions associated with grazing-avoidance. Given that  
13 grazing pressure has remained consistent throughout all years, these patterns suggest a  
14 combined effect of climate factors in driving these changes, rather than grazing alone.

#### 15 *Effect of climatic extremes on communities' functional composition*

16 Results from the models showed that the occurrence of prolonged climate extremes had a  
17 strong effect on the CWM rate of change of most of the traits considered, with grazed  
18 communities showing higher sensitivity, as indicated by the model's slope coefficients  
19 (supplementary Table S8 [**Supplementary Information**]). Climate extremes related to  
20 precipitation alteration tended to have the highest impact on trait change. Indeed, prolonged  
21 spring drought events (DRY.DAYS) promoted a shift towards a more drought-tolerant  
22 community in both grazed and ungrazed plots. When considering heatwaves duration  
23 (HEATWAVES), we found the opposite pattern in both communities, with species with an  
24 acquisitive strategy being favored against species with a more conservative strategy in years  
25 with higher HEATWAVES values. Since we initially considered both DRY.DAYS and

1 HEATWAVES as proxies of water shortage conditions, we expected to find a similar  
2 response in trait CWM annual change rate as a response to high DRY.DAYS and  
3 HEATWAVES values. On the contrary, our results suggest that grasslands may respond  
4 differently to prolonged dry spells and the occurrence of heatwaves. In our study,  
5 DRY.DAYS tended to be negatively correlated with HEATWAVES (supplementary Table  
6 S3 [**Supplementary Information**]), indicating that the years with higher heatwaves were not  
7 necessarily the ones with reduced precipitation. We hypothesized that in years with reduced  
8 precipitation (i.e., the ones with higher DRY.DAYS), species with a more conservative and  
9 drought tolerant strategy could better cope with reduced water availability, compared to  
10 species with a more acquisitive and drought avoidant strategy (French *et al.* 2019). This  
11 strategy seemed successful in years with higher HEATWAVES, when a higher water  
12 transport efficiency (i.e. high VLA - Scoffoni *et al.* 2016) could have helped to sustain the  
13 transpiration rates determined by high air temperature and VPD, but not in years with limited  
14 water availability, like the ones with higher DRY.DAYS. To support this hypothesis, forb  
15 species, which tended to have higher VLA than the other species (supplementary Table S1  
16 [**Supplementary Information**]), seemed favored in years with higher HEATWAVES,  
17 confirming their ability to grow under sufficient soil moisture even when moderate heatwaves  
18 occur (De Boeck *et al.* 2016; Corona-Lozada *et al.* 2019). In addition, the shift toward a more  
19 drought-tolerant community in years with higher DRY.DAYS was likely due to the decrease  
20 in annual forbs and N-fixers like *T. incarnatum*, coupled with an increase in perennial  
21 drought tolerant grasses (supplementary Fig. S1 - Fig. S2 [**Supplementary Information**])  
22 (Gargallo-Garriga *et al.* 2014).

23 Our findings suggest a joint effect of grazing and climate extremes in shaping the functional  
24 composition of the grassland community. Specifically, selective pressure from grazing seems  
25 to favor grazing-tolerant species in years with ample water availability throughout the

1 vegetative season (Zhang *et al.* 2023) while, on the other hand, an increase in drought events  
2 in recent years has promoted a change in community composition towards species with  
3 grazing-avoidance strategies. Indeed, the trait suites of grazing-avoidant species not only  
4 provide an advantage in the context of grazing but also offer enhanced drought tolerance (de  
5 Bello *et al.* 2006), underscoring the convergent selective pressures of grazing and aridity  
6 (Carmona *et al.* 2012). These results also suggest that the changes in functional composition  
7 will likely affect ecosystem processes due to the combined effect of grazing and climate  
8 extremes. In particular, carbon and nutrient cycling are strongly correlated to LES traits such  
9 as LMA and LNC (Cornwell *et al.* 2008), and our results suggest that the interplay between  
10 grazing alteration and prolonged drought could favor species with slower decomposition rate,  
11 thus decreasing soil C and N availability (Luo *et al.* 2022). However, nutrient cycling in  
12 grasslands is a complex mechanism (Luo *et al.* 2022; Vilonen *et al.* 2022), and future studies  
13 are needed to improve our ability to predict its future dynamics.

## 14 **Conclusions**

15 Our findings underscore the intertwined roles of grazing and climate extremes as driving  
16 forces in the long-term functional composition of the grassland community. Grazing pressure  
17 promoted species that adopt more acquisitive and fast-growing strategies, generally  
18 contributing to a functionally diverse community. However, when coupled with prolonged  
19 drought events, we found a noticeable shift in community composition towards species with  
20 more conservative strategies, showing both drought-tolerant and grazing-avoidant strategies.  
21 Conversely, the ungrazed community, characterized by a dominance of species with more  
22 conservative resource acquisition strategies exhibited greater stability in functional richness  
23 and divergence, alongside reduced sensitivity to climatic extremes.

1           Although our results support a generalizable understanding of plant community  
2 response to grazing and climate extremes based on differences in functional traits, we  
3 acknowledge that our dataset has potential limitations. Specifically, traits were measured on  
4 few individuals per species in a single year, due to technical and logistical constraints. While  
5 similar trait-base approaches have been widely utilized in assessing functional composition  
6 over time (e.g., Vandewalle et al. 2014), future studies would benefit from repeated  
7 functional traits measurements across multiple years and larger number of individuals per  
8 species. Such longitudinal sampling would better capture intraspecific variability and  
9 phenotypic plasticity (Mudrák *et al.* 2019), thereby providing a more comprehensive view on  
10 the dynamics of functional composition changes in grasslands and their implications for  
11 community resilience.

12

### 13 **Funding**

14 Project funded under the National Recovery and Resilience Plan (NRRP), Mission 4  
15 Component 2 Investment 1.4 - Call for tender No. 3138 of 16 December 2021, rectified by  
16 Decree n.3175 of 18 December 2021 of Italian Ministry of University and Research funded  
17 by the European Union – NextGenerationEU. Project code CN\_00000033, Concession  
18 Decree No. 1034 of 17 June 2022 adopted by the Italian Ministry of University and Research,  
19 CUP B63C22000650007", Project title "National Biodiversity Future Center - NBFC". ET is  
20 currently supported by Estonian Research Council grants (MOBERC100 and PRG2142). FP  
21 is currently supported by the funding PON Ricerca e Innovazione D.M. 1062/21– Contratti di  
22 ricerca, from the Italian Ministry of University (MUR).

23

### 24 **Acknowledgments**

1 NP, ET, FP and GB conceptualized the study and planned the data analysis. NP, ET, FP, GB,  
2 SM and EB measured the functional traits. SM, GB and EF provided the vegetation data. NP  
3 and ET performed the data analysis. NP, ET and FP wrote the first draft of the manuscript.  
4 EB, SM, AN, EF, CR, DR, LC and MC contributed to the critical revision of the manuscript  
5 to produce the final version.

6

### 7 **Conflict of Interest Statement**

8 The authors declare no competing interests.

9

### 10 **Data availability**

11 The datasets along with the R material needed to replicate the main analysis and figures in  
12 this manuscript are available in ‘figshare’ under embargo (private link:  
13 <https://figshare.com/s/63166078e4776aebe8e8> )

14

15

16

17

18

19

20

## 1 Literature Cited

- 2 **Anderegg WRL, Klein T, Bartlett M, et al. 2016.** Meta-analysis reveals that hydraulic traits  
3 explain cross-species patterns of drought-induced tree mortality across the globe.  
4 *Proceedings of the National Academy of Sciences* **113**: 5024–5029.
- 5 **Avolio ML, Forrestel EJ, Chang CC, La Pierre KJ, Burghardt KT, Smith MD. 2019.**  
6 Demystifying dominant species. *New Phytologist* **223**: 1106–1126.
- 7 **Bardgett RD, Bullock JM, Lavorel S, et al. 2021.** Combatting global grassland degradation.  
8 *Nature Reviews Earth & Environment* **2**: 720–735.
- 9 **Beccari E., Lanzaco PC, Salguero-Gómez R, Carmona CP. 2024.** Global Diversity in  
10 Mammalian Life Histories: Environmental Realms and Evolutionary Adaptations. :  
11 2023.06.29.546851.
- 12 **Beccari Eleonora, Pérez Carmona C, Tordoni E, et al. 2024.** Plant spectral diversity from  
13 high-resolution multispectral imagery detects functional diversity patterns in coastal dune  
14 communities. *Journal of Vegetation Science* **35**: e13239.
- 15 **de Bello F, Lepš J, Sebastià M-T. 2006.** Variations in species and functional plant diversity  
16 along climatic and grazing gradients. *Ecography* **29**: 801–810.
- 17 **Bengtsson J, Bullock JM, Egoh B, et al. 2019.** Grasslands—more important for ecosystem  
18 services than you might think. *Ecosphere* **10**: e02582.
- 19 **Carmona CP, Azcárate FM, de Bello F, Ollero HS, Lepš J, Peco B. 2012.** Taxonomical  
20 and functional diversity turnover in Mediterranean grasslands: interactions between grazing,  
21 habitat type and rainfall (M Cadotte, Ed.). *Journal of Applied Ecology* **49**: 1084–1093.
- 22 **Carmona CP, Bello F de, Mason NWH, Lepš J. 2016.** Traits Without Borders: Integrating  
23 Functional Diversity Across Scales. *Trends in Ecology & Evolution* **31**: 382–394.
- 24 **Carmona CP, Bello F, Mason NWH, Lepš J. 2019.** Trait probability density (TPD):  
25 measuring functional diversity across scales based on TPD with R. *Ecology* **100**.
- 26 **Carmona CP, Bueno CG, Toussaint A, et al. 2021.** Fine-root traits in the global spectrum  
27 of plant form and function. *Nature* **597**: 683–687.
- 28 **Carmona CP, Pavanetto N, Puglielli G. 2024.** funspace: An R package to build, analyse  
29 and plot functional trait spaces. *Diversity and Distributions* **30**: e13820.
- 30 **Cornwell WK, Cornelissen JHC, Amatangelo K, et al. 2008.** Plant species traits are the  
31 predominant control on litter decomposition rates within biomes worldwide. *Ecology Letters*  
32 **11**: 1065–1071.
- 33 **Corona-Lozada MC, Morin S, Choler P. 2019.** Drought offsets the positive effect of  
34 summer heat waves on the canopy greenness of mountain grasslands. *Agricultural and*  
35 *Forest Meteorology* **276–277**: 107617.

- 1 **De Boeck HJ, Bassin S, Verlinden M, Zeiter M, Hiltbrunner E. 2016.** Simulated heat  
2 waves affected alpine grassland only in combination with drought. *The New Phytologist* **209**:  
3 531–541.
- 4 **Dengler J, Biurrun I, Boch S, Dembicz I, Török P. 2020.** Grasslands of the Palaeartic  
5 Biogeographic Realm: Introduction and Synthesis In: *Encyclopedia of the World's Biomes*.  
6 Elsevier, 617–637.
- 7 **Díaz S, Kattge J, Cornelissen JHC, et al. 2016.** The global spectrum of plant form and  
8 function. *Nature* **529**: 167–171.
- 9 **Díaz S, Lavorel S, McINTYRE S, et al. 2007.** Plant trait responses to grazing? a global  
10 synthesis. *Global Change Biology* **13**: 313–341.
- 11 **Dormann CF, Elith J, Bacher S, et al. 2013.** Collinearity: a review of methods to deal with  
12 it and a simulation study evaluating their performance. *Ecography* **36**: 27–46.
- 13 **Dukes JS, Chiariello NR, Cleland EE, et al. 2005.** Responses of Grassland Production to  
14 Single and Multiple Global Environmental Changes. *PLOS Biology* **3**: e319.
- 15 **Elst EM, De Boeck HJ, Vanmaele L, Verlinden M, Dhliwayo P, Nijs I. 2017.** Impact of  
16 climate extremes modulated by species characteristics and richness. *Perspectives in Plant*  
17 *Ecology, Evolution and Systematics* **24**: 80–92.
- 18 **Evans SE, Byrne KM, Lauenroth WK, Burke IC. 2011.** Defining the limit to resistance in  
19 a drought-tolerant grassland: long-term severe drought significantly reduces the dominant  
20 species and increases ruderals. *Journal of Ecology* **99**: 1500–1507.
- 21 **Farquhar GD, Ehleringer JR, Hubick KT. 1989.** Carbon Isotope Discrimination and  
22 Photosynthesis. *Annual Review of Plant Biology* **40**: 503–537.
- 23 **French K, Jansens IB, Ashcroft MB, Ecroyd H, Robinson SA. 2019.** High tolerance of  
24 repeated heatwaves in Australian native plants. *Austral Ecology* **44**: 597–608.
- 25 **Gargallo-Garriga A, Sardans J, Pérez-Trujillo M, et al. 2014.** Opposite metabolic  
26 responses of shoots and roots to drought. *Scientific Reports* **4**: 6829.
- 27 **Gleason SM, Westoby M, Jansen S, et al. 2016.** Weak tradeoff between xylem safety and  
28 xylem-specific hydraulic efficiency across the world's woody plant species. *New Phytologist*  
29 **209**: 123–136.
- 30 **Griffin-Nolan RJ, Blumenthal DM, Collins SL, et al. 2019.** Shifts in plant functional  
31 composition following long-term drought in grasslands. *Journal of Ecology* **107**: 2133–2148.
- 32 **Griffin-Nolan RJ, Slette IJ, Knapp AK. 2021.** Deconstructing precipitation variability:  
33 Rainfall event size and timing uniquely alter ecosystem dynamics. *Journal of Ecology* **109**:  
34 3356–3369.
- 35 **Grime JP. 1998.** Benefits of plant diversity to ecosystems: immediate, filter and founder  
36 effects. *Journal of Ecology* **86**: 902–910.

- 1 **Hallett LM, Hsu JS, Cleland EE, et al. 2014.** Biotic mechanisms of community stability  
2 shift along a precipitation gradient. *Ecology* **95**: 1693–1700.
- 3 **Hallett LM, Jones SK, MacDonald AAM, et al. 2016.** codyn: An r package of community  
4 dynamics metrics. *Methods in Ecology and Evolution* **7**: 1146–1151.
- 5 **Harrison SP, LaForgia ML, Latimer AM. 2018.** Climate-driven diversity change in annual  
6 grasslands: Drought plus deluge does not equal normal. *Global Change Biology* **24**: 1782–  
7 1792.
- 8 **Hayes GF, Holl KD. 2003.** Cattle Grazing Impacts on Annual Forbs and Vegetation  
9 Composition of Mesic Grasslands in California. *Conservation Biology* **17**: 1694–1702.
- 10 **Hoover DL, Knapp AK, Smith MD. 2014.** Resistance and resilience of a grassland  
11 ecosystem to climate extremes. *Ecology* **95**: 2646–2656.
- 12 **Kassambara A. 2023.** rstatix: Pipe-Friendly Framework for Basic Statistical Tests.
- 13 **Komac B, Pladevall C, Domènech M, Fanlo R. 2015.** Functional diversity and grazing  
14 intensity in sub-alpine and alpine grasslands in Andorra. *Applied Vegetation Science* **18**: 75–  
15 85.
- 16 **Kramp RE, Liancourt P, Herberich MM, et al. 2022.** Functional traits and their plasticity  
17 shift from tolerant to avoidant under extreme drought. *Ecology* **103**: e3826.
- 18 **Kurze S, Bilton MC, Álvarez-Cansino L, et al. 2021.** Evaluating grazing response  
19 strategies in winter annuals: A multi-trait approach. *Journal of Ecology* **109**: 3074–3086.
- 20 **Lavorel S, Díaz S, Cornelissen JHC, et al. 2007.** Plant Functional Types: Are We Getting  
21 Any Closer to the Holy Grail? In: Canadell JG, Pataki DE, Pitelka LF, eds. *Terrestrial*  
22 *Ecosystems in a Changing World*. Berlin, Heidelberg: Springer, 149–164.
- 23 **Liira J, Schmidt T, Aavik T, et al. 2008.** Plant functional group composition and large-scale  
24 species richness in European agricultural landscapes. *Journal of Vegetation Science* **19**: 3–14.
- 25 **Lin X, Zhao H, Zhang S, et al. 2024.** Global response of different types of grasslands to  
26 precipitation and grazing, especially belowground biomass. *Agriculture, Ecosystems &*  
27 *Environment* **363**: 108852.
- 28 **Lionello P, Scarascia L. 2018.** The relation between climate change in the Mediterranean  
29 region and global warming. *Regional Environmental Change* **18**: 1481–1493.
- 30 **Luo W, Zuo X, Griffin-Nolan RJ, et al. 2019.** Long term experimental drought alters  
31 community plant trait variation, not trait means, across three semiarid grasslands. *Plant and*  
32 *Soil* **442**: 343–353.
- 33 **Luo W, Zuo X, Griffin-Nolan RJ, et al. 2022.** Chronic and intense droughts differentially  
34 influence grassland carbon-nutrient dynamics along a natural aridity gradient. *Plant and Soil*  
35 **473**: 137–148.
- 36 **Maccherini S, Bacaro G, Marignani M. 2018.** Beneficial effects of restoration practices can  
37 be thwarted by climate extremes. *Science of The Total Environment* **626**: 851–859.

- 1 **Maccherini S, Santi E. 2012.** Long-term experimental restoration in a calcareous grassland:  
2 Identifying the most effective restoration strategies. *Biological Conservation* **146**: 123–135.
- 3 **Magurran AE, Dornelas M, Moyes F, Henderson PA. 2019.** Temporal  $\beta$  diversity—A  
4 macroecological perspective (D Storch, Ed.). *Global Ecology and Biogeography* **28**: 1949–  
5 1960.
- 6 **Maitner BS, Halbritter AH, Telford RJ, et al. 2023.** Bootstrapping outperforms  
7 community-weighted approaches for estimating the shapes of phenotypic distributions.  
8 *Methods in Ecology and Evolution* **14**: 2592–2610.
- 9 **Martínez-Villa JA, Durán SM, Enquist BJ, Duque A, Messier C, Paquette A. 2024.**  
10 Temporal shifts in the functional composition of Andean forests at different elevations are  
11 driven by climate change. *Global Ecology and Biogeography* **33**: 85–99.
- 12 **Mouillot D, Villéger S, Scherer-Lorenzen M, Mason NWH. 2011.** Functional Structure of  
13 Biological Communities Predicts Ecosystem Multifunctionality. *PLOS ONE* **6**: e17476.
- 14 **Mudrák O, Doležal J, Vítová A, Lepš J. 2019.** Variation in plant functional traits is best  
15 explained by the species identity: Stability of trait-based species ranking across meadow  
16 management regimes (M Weiser, Ed.). *Functional Ecology* **33**: 746–755.
- 17 **Myhre G, Alterskjær K, Stjern CW, et al. 2019.** Frequency of extreme precipitation  
18 increases extensively with event rareness under global warming. *Scientific Reports* **9**: 16063.
- 19 **Nardini A, Luglio J. 2014.** Leaf hydraulic capacity and drought vulnerability: possible trade-  
20 offs and correlations with climate across three major biomes. *Functional Ecology* **28**: 810–  
21 818.
- 22 **Pavanetto N, Carmona CP, Laanisto L, Niinemets Ü, Puglielli G. 2024.** Trait dimensions  
23 of abiotic stress tolerance in woody plants of the Northern Hemisphere. *Global Ecology and*  
24 *Biogeography* **33**: 272–285.
- 25 **Pavanetto N, Niinemets Ü, Rueda M, Puglielli G. 2024.** Macroecology of Abiotic Stress  
26 Tolerance in Woody Plants of the Northern Hemisphere: Tolerance Biomes and  
27 Polytolerance Hotspots. *Ecology Letters* **27**: e70016.
- 28 **Peñuelas J, Sardans J, Filella I, et al. 2017.** Impacts of Global Change on Mediterranean  
29 Forests and Their Services. *Forests* **8**: 463.
- 30 **Pérez-Harguindeguy N, Díaz S, Garnier E, et al. 2013.** New handbook for standardised  
31 measurement of plant functional traits worldwide. *Australian Journal of Botany* **61**: 167–234.
- 32 **Petruzzellis F, Tordoni E, Di Bonaventura A, et al. 2022.** Turgor loss point and  
33 vulnerability to xylem embolism predict species-specific risk of drought-induced decline of  
34 urban trees. *Plant Biology* **24**: 1198–1207.
- 35 **Petruzzellis F, Tordoni E, Tomasella M, et al. 2021.** Functional differentiation of invasive  
36 and native plants along a leaf efficiency/safety trade-off. *Environmental and Experimental*  
37 *Botany* **188**: 104518.

- 1 **Poniatowski D, Hertenstein F, Raude N, Gottbehüt K, Nickel H, Fartmann T. 2018.** The  
2 invasion of *Bromus erectus* alters species diversity of vascular plants and leafhoppers in  
3 calcareous grasslands (A Stewart, Ed.). *Insect Conservation and Diversity* **11**: 578–586.
- 4 **Pörtner H-O, Roberts DC, Adams H, et al. 2022.** Climate Change 2022: Impacts,  
5 Adaptation and Vulnerability.
- 6 **Prangel E, Reitalu T, Neuenkamp L, et al. 2024.** Restoration of semi-natural grasslands  
7 boosts biodiversity and re-creates hotspots for ecosystem services. *Agriculture, Ecosystems*  
8 *& Environment* **374**: 109139.
- 9 **Puglielli G, Bricca A, Chelli S, et al. 2024.** Intraspecific variability of leaf form and function  
10 across habitat types. *Ecology Letters* **27**: e14396.
- 11 **Puglielli G, Pavanetto N, Laanisto L. 2022.** Towards a “periodic table” of abiotic stress  
12 tolerance strategies of woody plants. *Flora* **292**: 152089.
- 13 **Qu L, Dong G, De Boeck HJ, et al. 2020.** Joint forcing by heat waves and mowing poses a  
14 threat to grassland ecosystems: Evidence from a manipulative experiment. *Land Degradation*  
15 *& Development* **31**: 785–800.
- 16 **Reich PB. 2014.** The world-wide ‘fast-slow’ plant economics spectrum: a traits manifesto (H  
17 Cornelissen, Ed.). *Journal of Ecology* **102**: 275–301.
- 18 **Revell LJ. 2009.** Size-Correction and Principal Components for Interspecific Comparative  
19 Studies. *Evolution* **63**: 3258–3268.
- 20 **Rota C, Manzano P, Carmona CP, Malo JE, Peco B. 2017.** Plant community assembly in  
21 Mediterranean grasslands: understanding the interplay between grazing and spatio-temporal  
22 water availability (S Bartha, Ed.). *Journal of Vegetation Science* **28**: 149–159.
- 23 **Sack L, Scoffoni C, John GP, et al. 2013.** How do leaf veins influence the worldwide leaf  
24 economic spectrum? Review and synthesis. *Journal of Experimental Botany* **64**: 4053–4080.
- 25 **Sandel B, Goldstein LJ, Kraft NJB, et al. 2010.** Contrasting trait responses in plant  
26 communities to experimental and geographic variation in precipitation. *New Phytologist* **188**:  
27 565–575.
- 28 **Sasaki T, Lauenroth WK. 2011.** Dominant species, rather than diversity, regulates temporal  
29 stability of plant communities. *Oecologia* **166**: 761–768.
- 30 **Scoffoni C, Chatelet DS, Pasquet-kok J, et al. 2016.** Hydraulic basis for the evolution of  
31 photosynthetic productivity. *Nature Plants* **2**: 1–8.
- 32 **Šimová I, Violle C, Svenning J-C, et al. 2018.** Spatial patterns and climate relationships of  
33 major plant traits in the New World differ between woody and herbaceous species. *Journal*  
34 *of Biogeography* **45**: 895–916.
- 35 **Smith MD. 2011.** An ecological perspective on extreme climatic events: a synthetic  
36 definition and framework to guide future research. *Journal of Ecology* **99**: 656–663.

- 1 **Stampfli A, Bloor JMG, Fischer M, Zeiter M. 2018.** High land-use intensity exacerbates  
2 shifts in grassland vegetation composition after severe experimental drought. *Global Change*  
3 *Biology* **24**: 2021–2034.
- 4 **Strona G, Nappo D, Boccacci F, Fattorini S, San-Miguel-Ayanz J. 2014.** A fast and  
5 unbiased procedure to randomize ecological binary matrices with fixed row and column  
6 totals. *Nature Communications* **5**: 4114.
- 7 **Sun S, Jung E-Y, Gaviria J, Engelbrecht BMJ. 2020.** Drought survival is positively  
8 associated with high turgor loss points in temperate perennial grassland species. *Functional*  
9 *Ecology* **34**: 788–798.
- 10 **Telford RJ, Halbritter AH, Maitner BS. 2023.** traitstrap: Bootstrap Trait Values to  
11 Calculate Moments.
- 12 **Teuling AJ, Seneviratne SI, Stöckli R, et al. 2010.** Contrasting response of European forest  
13 and grassland energy exchange to heatwaves. *Nature Geoscience* **3**: 722–727.
- 14 **Tordoni E, Petruzzellis F, Nardini A, Bacaro G. 2020.** Functional Divergence Drives  
15 Invasibility of Plant Communities at the Edges of a Resource Availability Gradient. *Diversity*  
16 **12**: 148.
- 17 **Tordoni E, Petruzzellis F, Nardini A, Savi T, Bacaro G. 2019.** Make it simpler: Alien  
18 species decrease functional diversity of coastal plant communities (Z Botta □ Dukát, Ed.).  
19 *Journal of Vegetation Science* **30**: 498–509.
- 20 **Trugman AT, Anderegg LDL, Shaw JD, Anderegg WRL. 2020.** Trait velocities reveal  
21 that mortality has driven widespread coordinated shifts in forest hydraulic trait composition.  
22 *Proceedings of the National Academy of Sciences* **117**: 8532–8538.
- 23 **Vandewalle M, Purschke O, Bello F de, et al. 2014.** Functional responses of plant  
24 communities to management, landscape and historical factors in semi-natural grasslands.  
25 *Journal of Vegetation Science* **25**: 750–759.
- 26 **Vilonen LL, Blair J, Trivedi P, Zeglin L, Smith MD. 2022.** Limited legacy effects of  
27 extreme multiyear drought on carbon and nitrogen cycling in a mesic grassland. *Elementa:*  
28 *Science of the Anthropocene* **10**: 000093.
- 29 **Vitra A, Deléglise C, Meisser M, et al. 2019.** Responses of plant leaf economic and  
30 hydraulic traits mediate the effects of early- and late-season drought on grassland  
31 productivity. *AoB PLANTS* **11**: plz023.
- 32 **Wang X, Chen S, Yang X, et al. 2024.** Adaptation mechanisms of leaf vein traits to drought  
33 in grassland plants. *Science of The Total Environment* **917**: 170224.
- 34 **Westoby M, Falster DS, Moles AT, Vesk PA, Wright IJ. 2002.** Plant Ecological  
35 Strategies: Some Leading Dimensions of Variation Between Species. *Annual Review of*  
36 *Ecology and Systematics* **33**: 125–159.
- 37 **Wright IJ, Reich PB, Westoby M, et al. 2004.** The worldwide leaf economics spectrum.  
38 *Nature* **428**: 821–827.

- 1 **Yan C-L, Ni M-Y, Cao K-F, Zhu S-D. 2020.** Leaf hydraulic safety margin and safety–  
2 efficiency trade-off across angiosperm woody species. *Biology Letters* **16**: 20200456.
- 3 **Zhang X, Alexander L, Hegerl GC, et al. 2011.** Indices for monitoring changes in extremes  
4 based on daily temperature and precipitation data. *WIREs Climate Change* **2**: 851–870.
- 5 **Zhang M, Delgado-Baquerizo M, Li G, et al. 2023.** Experimental impacts of grazing on  
6 grassland biodiversity and function are explained by aridity. *Nature Communications* **14**:  
7 5040.
- 8 **Zhu S-D, Chen Y-J, Ye Q, et al. 2018.** Leaf turgor loss point is correlated with drought  
9 tolerance and leaf carbon economics traits. *Tree Physiology* **38**: 658–663.

10

11

12

13

14

15

16

17

18

19

20

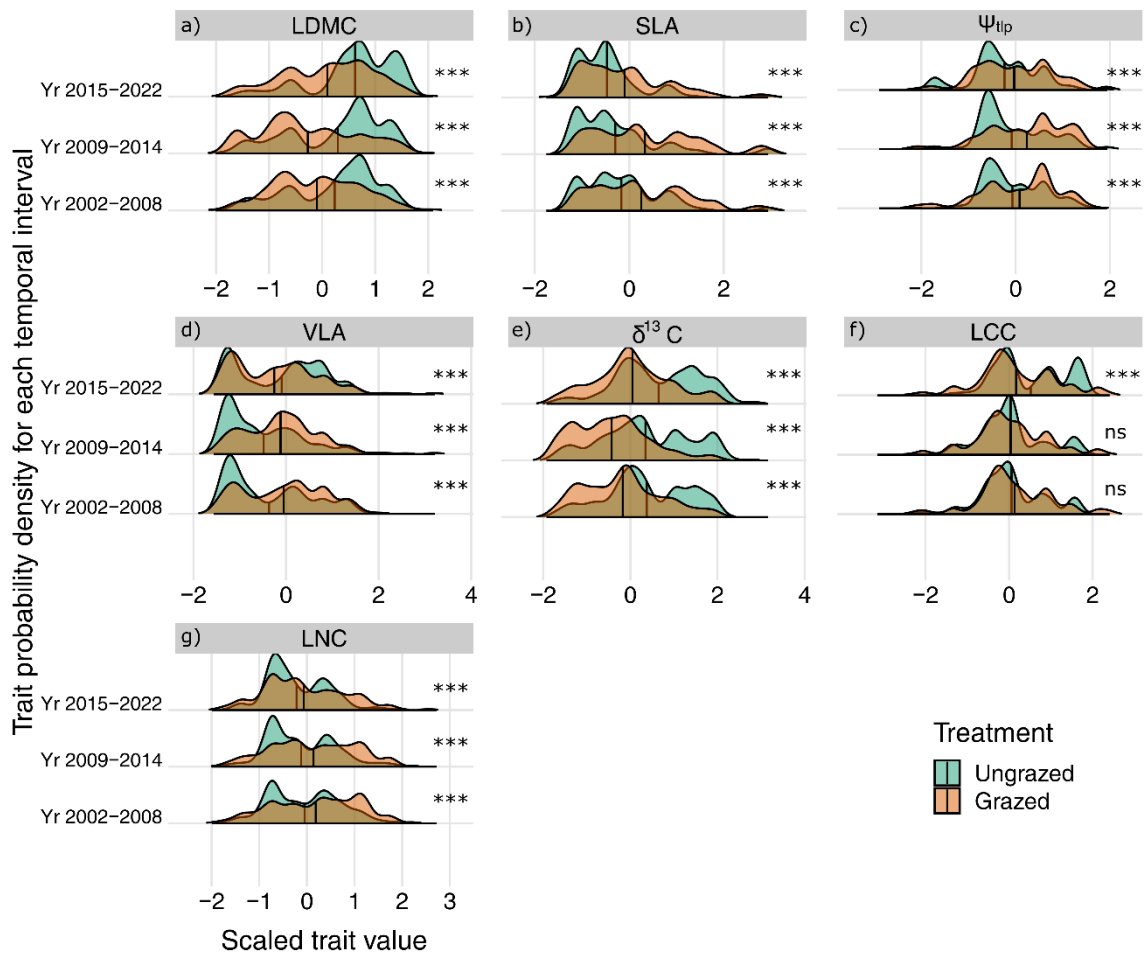
21

22

23

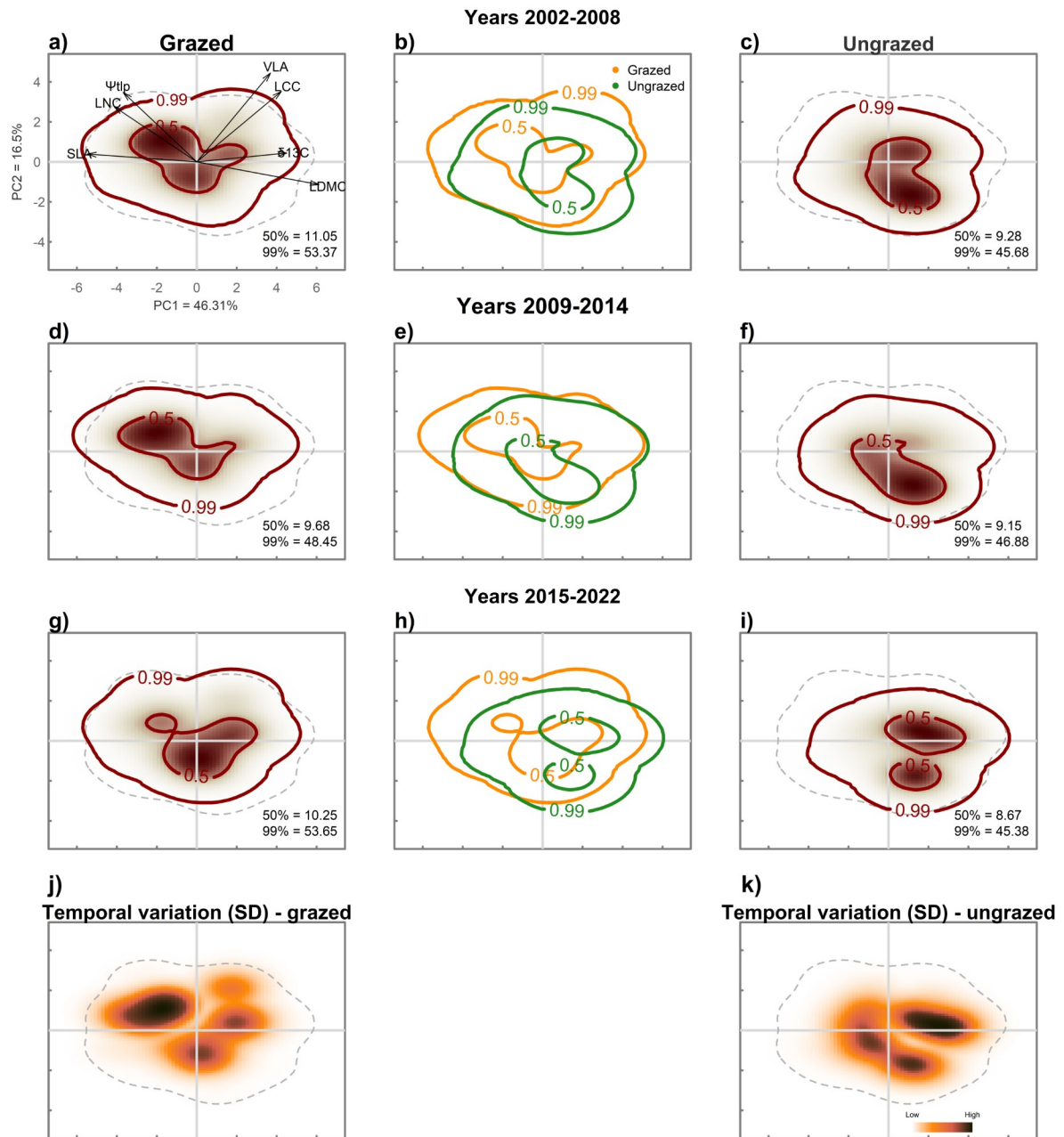
24

# 1 Figures



2

3 **Fig. 1.** Probability density distributions of each trait in grazed (orange) and ungrazed (green)  
 4 community for each temporal interval, obtained after the non-parametric bootstrapping of  
 5 trait data. Trait probability densities were estimated separately for each treatment. The  
 6 vertical black lines within the density plots represent the mean value of the distribution. (a)  
 7 leaf dry matter content (LDMC), (b) specific leaf area (SLA), (c) water potential at turgor  
 8 loss point ( $\Psi_{tlp}$ ), (d) minor vein length per unit area (VLA), (e) carbon stable isotope  
 9 composition ( $\delta^{13}C$ ), (f) leaf carbon content (LCC), (g) leaf nitrogen content (LNC). P-values  
 10 of Dunn's test with Holm-Bonferroni correction for multiple comparison testing for the  
 11 differences in the distribution of each trait values between grazed and ungrazed community  
 12 for each temporal interval are reported (ns: not significant; \*\*:  $p < 0.01$ ; \*\*\*:  $p < 0.001$ ).



1

2 **Fig. 2.** Probabilistic distribution of trait combinations in the functional trait space defined by  
 3 the PCA in the grazed (panels a,d,g) and ungrazed panels c, f and i) community for each  
 4 temporal interval. The colours indicate the probability densities, from high (dark brown) to  
 5 low (light brown) probability. Contour lines indicate the 0.99, 0.50 and 0.25 quantile of the  
 6 total probability distribution. Functional richness (FRic) values estimated considering 0.50  
 7 and 0.99 quantile for each community and temporal interval are reported. Central panels (b, e,

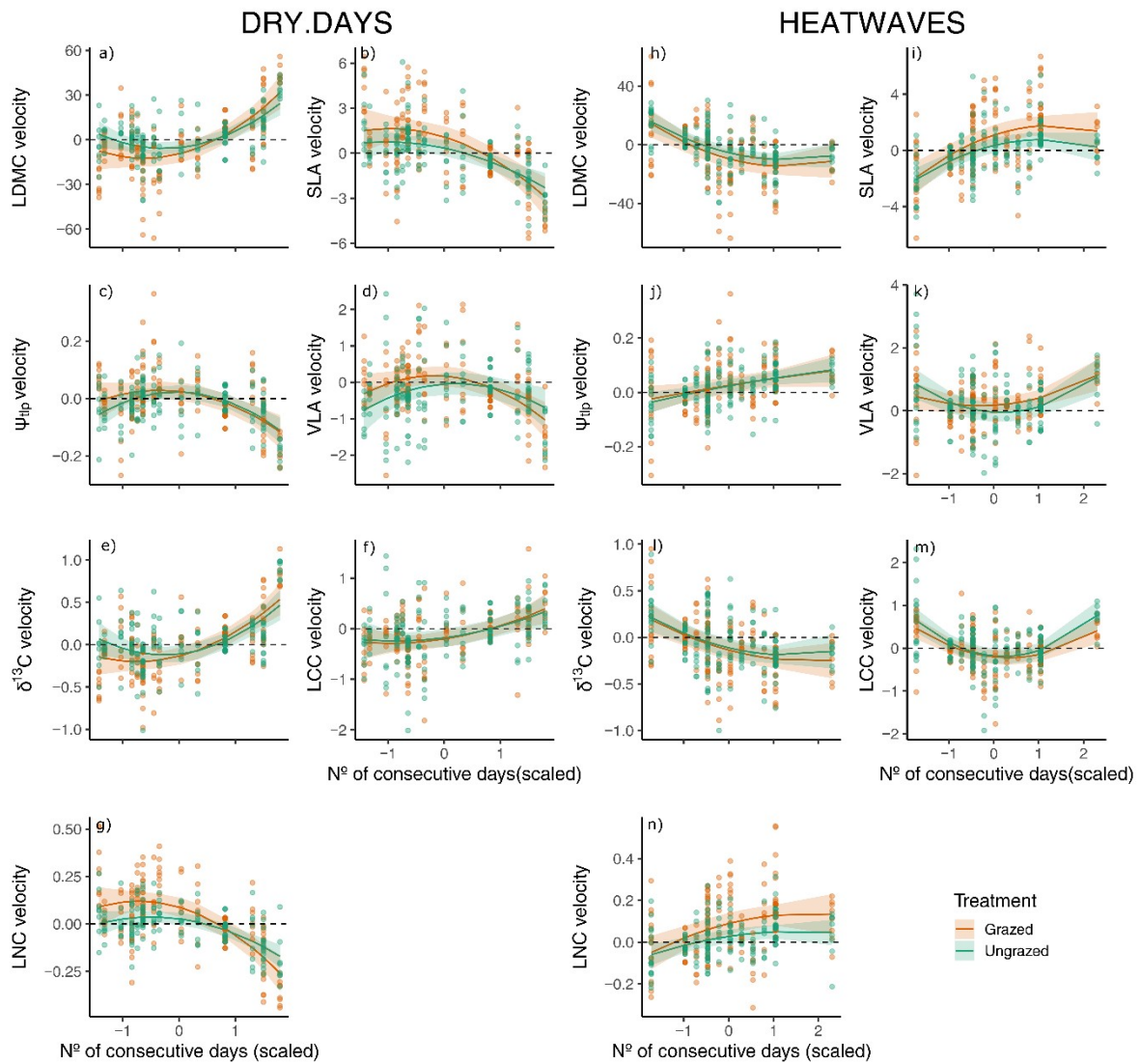
1 h) show the overlap between the two communities in each temporal interval (orange: grazed  
2 community; green: ungrazed community). Bottom panels (j, k) shows the total temporal  
3 variation of functional structure (expressed as standard deviation) for the grazed (i) and  
4 ungrazed (k) community, from higher (dark brown) to lower (light brown) values of temporal  
5 variation. Variables are: leaf dry matter content (LDMC), specific leaf area (SLA), water  
6 potential at turgor loss point ( $\Psi_{tlp}$ ), minor vein length per unit area (VLA), carbon stable  
7 isotope composition ( $\delta^{13}C$ ), leaf carbon content (LCC), leaf nitrogen content (LNC).

8

9

10

11



1

2 **Fig. 3.** Relationships between prolonged periods of drought and heat and annual rate of  
 3 change (i.e. velocity) of (community weighted) trait means for grazed and ungrazed  
 4 community. Trait velocity of zero represent no change over time, while positive and negative  
 5 values represent an increase vs decrease in trait community value respectively. Linear mixed  
 6 models were fitted separately for the two treatments. Only significant relationships are  
 7 reported (see Table S8 [Supporting Information] for models' coefficients and statistics). Each  
 8 point represents the trait velocity for each plot in a given year, for grazed (orange) and  
 9 ungrazed (green) community. Ribbons represent the 95% confidence intervals. Climatic  
 10 variables were scaled to mean zero and variance one.(a,h) leaf dry matter content (LDMC),

- 1 (b,i) specific leaf area (SLA), (c,j) water potential at turgor loss point ( $\Psi_{tp}$ ), (d,k) minor vein
- 2 length per unit area (VLA), (e,l) carbon stable isotope composition ( $\delta^{13}C$ ), (f,m) leaf carbon
- 3 content (LCC), (g,n) leaf nitrogen content (LNC). Climate variables are: Maximum
- 4 consecutive number of consecutive days with precipitation  $<1$  mm in a given year
- 5 (DRY.DAYS) and maximum consecutive number of annual days with  $T > 30^{\circ} C$  in a given
- 6 year (HEATWAVES).