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## RESEARCH ARTICLE



## Climate-induced plasticity in leaf traits of riparian plants

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## Abstract

**Aim:** Leaf litter inputs from riparian vegetation and its decomposition play a key role in energy and nutrient transfer in many stream ecosystems. Instream leaf litter decomposition is driven by both leaf traits and environmental conditions. Therefore, understanding and predicting leaf trait variation under current environmental changes and their putative interactive effects on stream food webs is a critical challenge. Most studies have focussed on the assumed higher interspecific leaf trait variability, with little research addressing an intraspecific perspective.

Location: Andalusia, Spain.

**Methods:** We assessed the relative effects of climate and soil conditions on the plasticity of leaf traits of four common woody riparian species in permanent low-order Mediterranean streams across a wide aridity gradient. We used a space-for-time substitution approach to predict leaf trait changes and consequences for stream food webs in a future climate change scenario.

**Results:** Overall, we found that aridity had a major influence on leaf trait plasticity but with opposite patterns depending on plant functional type, although soil was the strongest predictor in some cases. Results indicated that leaf quality—linked to palatability and decomposability—of *Alnus glutinosa*, *Salix atrocinerea* and *Rubus ulmifolius* (deciduous/semi-deciduous) will decrease with forecasted aridification, whereas the palatability of the evergreen *Nerium oleander* will increase. We observed higher trait plasticity than interspecific variation for leaf P, Ca and Mg concentrations and C:P ratio.

**Main conclusions:** Our findings suggest a decrease of intraspecific leaf quality in riparian deciduous species with global warming in a relatively short term. In a longer term, this may merge with the forecasted dieback of deciduous species in riparian corridors of temperate climate zones. These changes have the potential to significantly impair ecosystem functioning of Mediterranean mountain streams currently under deciduous gallery forests.

#### KEYWORDS

Aridification, deciduous, evergreen, instream decomposition, litter quality, soil, space-for-time substitution

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## 1 | INTRODUCTION

The warming of the Earth system is unequivocal (IPCC, 2021). Globally, precipitation is also predicted to increase in the long term (Hewitson et al., 2015). However, forecasts in the Mediterranean basin point to a precipitation decrease of around 34%, along with a temperature increase of *ca*. 5°C for the period 2000–2099 (Harris et al., 2013). As a result, this region will face a climate much drier and hotter than at present, especially during warm seasons (Giorgi & Lionello, 2008), with direct effects on hydrologic regimes (Nohara et al., 2006; Vicente-Serrano et al., 2014) and soil moisture (Manabe et al., 2004). These changes may alter the functioning and structure of plant communities (e.g. Carnicer et al., 2011; Trivedi et al., 2008; Vicente-Serrano et al., 2012).

Small streams flowing through forested areas can be especially susceptible to climate change-induced alterations in plant communities, owing to their high dependence on organic matter inputs from the riparian vegetation, i.e. leaf litter (Wallace et al., 2015). Instream decomposition of leaf litter is a crucial ecosystem process, involving the cycling of nutrients and fuelling stream secondary production (Marks, 2019). The rate at which leaf litter decomposes and is incorporated into food webs highly depends on its quality, which fundamentally depends on after-life persistent traits (Graca & Cressa, 2010; Graça et al., 2001; Zhang et al., 2019). Thus, ecosystem functioning can be significantly altered if leaf litter inputs to streams experience physical and chemical changes (e.g. Casas et al., 2013; del Campo et al., 2021; López-Rojo et al., 2019). These changes can be interspecific, e.g. resulting from the forecasted substitution of deciduous by evergreen species (Kominoski et al., 2013; Salinas et al., 2018) and/or the decline of key plant species populations (e.g. alder; Alonso et al., 2021; Rubio-Ríos et al., 2021). Moreover, given that leaf traits are highly responsive to environmental changes (Heilmeier, 2019; Soudzilovskaia et al., 2013), intraspecific changes may also occur, e.g. due to genetic variability (Crutsinger et al., 2014; LeRoy et al., 2012) or phenotypic plasticity (Graca & Poquet, 2014; Henn et al., 2018; Jung et al., 2014).

Such relationship between leaf traits and the environment has been a recurrent theme of the study (e.g. Ordoñez et al., 2009; Read et al., 2014; Reich & Oleksyn, 2004). However, although recent results indicate that intraspecific variation may represent up to *ca*. 30% of total functional trait variability in plant communities (Albert et al., 2010; Siefert et al., 2015), most studies have focussed on the often assumed higher interspecific variability of many leaf traits (e.g. Hulshof & Swenson, 2010; Wright et al., 2004).

High rates of plasticity in leaf traits are expected in species distributed across ample environmental gradients (Cordell et al., 1998; Fajardo & Piper, 2011; Umaña & Swenson, 2019), as increases in niche breadth allow plants to respond to variation in climatic and other environmental conditions (Henn et al., 2018), whereas nearby individuals may share biotic and abiotic pressures and have close genetic relationships. Warming and reduced rainfall, i.e. increasing aridity, are usually reported to promote the production of thicker and smaller leaves (Wright et al., 2004)—in order to improve their water use efficiency and to increase their leaf life span—with low nutrient concentrations (Reich & Oleksyn, 2004). Such plasticity in important traits can, in turn, affect the palatability and decomposability of leaves, i.e. their acceptability and easiness to be consumed, along environmental gradients (Boyero et al., 2017; Graça & Poquet, 2014; Lecerf & Chauvet, 2008; LeRoy et al., 2007). Understanding how individual species traits, or their syndromes, are modulated by climatic or other environmental characteristics could allow us to refine predictions of potential effects on stream ecosystem functioning, both in green (based on primary production) and brown (based on detritus) food webs, in the face of climate change (Kominoski et al., 2021).

Here, using a 'space-for-time' (SFT) substitution approach (Blois et al., 2013; Pickett, 1989), we investigated how climate change might affect leaf quality, focusing on after-life traits affecting leaf decomposition. The SFT substitution approach is a useful tool to anticipate changes taking advantage of natural gradients (Fukami & Wardle, 2005); in the present study, a natural aridity gradient represents the forecasted aridification of the Mediterranean basin (Seager et al., 2014). We assessed plasticity in leaf traits of four common riparian species, with contrasting functional traits, in permanent low-order streams [Alnus glutinosa (L.) Gaertn., Salix atrocinerea Brot., Rubus ulmifolius Schott and Nerium oleander L.], extrapolating their possible variation in the forecasted climatic scenarios from that observed across a wide environmental gradient studied within a relatively small region. Using the same species along many areas differing in environmental conditions allowed us to control for species-specific traits, but not to assess the amount of trait variability due to genetic variability.

Given the high responsiveness of leaves to climate changes (Heilmeier, 2019; Soudzilovskaia et al., 2013) and the high water and nutrient availability in riparian soils of permanent streams (Naiman & Decamps, 1997), we hypothesize that (1) climate will exert a higher influence on leaf trait plasticity of the studied species compared to soil variables. We (2) expect a general trend of decreasing leaf quality—i.e. lower nutrient concentration, higher toughness—with the forecasted aridification (increasing temperature and decreasing precipitation) (Reich & Oleksyn, 2004). However, we also expect that the strength of the effects will vary among different species, as they belong to different functional groups (i.e. C allocation and/or N-fixing) and therefore have low similarity in their leaf traits (Salinas et al., 2018). Thus, we also hypothesize that (3) trait plasticity will be relatively low compared to interspecific variation.

## 2 | MATERIALS AND METHODS

### 2.1 | Area of study and selected plant species

Our study was conducted during summer 2013 in the riverbanks of 34 headwater streams with permanent flows distributed across nine natural protected areas (considered as pristine) located in Andalusia (south of the Iberian Peninsula), covering *ca.* 88,000 km<sup>2</sup>. These

locations represent a wide climatic gradient within the context of a Mediterranean-type climate and possess a considerable lithological and topographical heterogeneity (Figure 1). The present (mean annual temperature range 10.8-17.4°C; mean annual precipitation range 261-845 mm; Table S1) and the projected climatic gradient studied (by the end of the 21st century) covers from arid to humid conditions according to the Emberger's bioclimatic coefficient (Table 1, Figure 2). This embraces the forecasted aridification, i.e. warming (mean temperature rise of 2-4°C) and reduction of precipitation (mean precipitation decrease of 10-40%), for the Mediterranean region (Seager et al., 2014), as a consequence of climate change towards the year 2100 (reviewed by Giorgi & Lionello, 2008).

We selected four abundant riparian plant species which represent different functional groups featuring different characteristics, including two deciduous riparian trees: black alder-Alnus glutinosa (an N-fixer), and grey willow, Salix atrocinerea; one semi-deciduous shrub: blackberry, Rubus ulmifolius; and one evergreen shrub: oleander. Nerium oleander, also known as laurel rose. Leaves of these species collected (June-July 2013) from each sampling sites were present (Table 1) from robust, well-grown and totally unshaded plants distanced from the stream by a maximum of 6 m. Those leaves directly exposed to sun light and without herbivory or pathogen symptoms were selected (Cornelissen et al., 2003). In each stream

and for each species, we collected 102 leaves from six individuals (17 leaves per individual) randomly distributed on both stream sides along a 100 m stream reach. Leaves were air-dried at room temperature (20-23°C) for one week and stored in darkness in paper bags until processed. At each stream, the cover of each species was estimated using the Domin-Krajina scale of cover and abundance (Kent & Coker, 1992) in six plots (36 m<sup>2</sup> each) randomly distributed in both stream sides-three plots per side arranged from the edge of the wetted channel-along a 100 m stream reach (Salinas et al., 2018).

#### **Environmental variables** 2.2

Thirty-two environmental variables (altitude, 20 climatic and 11 edaphic; Table 1 and S1) were selected as potential predictors of leaf trait plasticity. Altitude was obtained in situ using a portable GPS. Historical (monthly average for the years 1970-2000) values of bioclimatic variables (spatial resolution of 30 seconds, i.e.  $\sim 1 \times 1$  km) recorded along the last period with available climatic data were obtained from the WorldClim database (Table S1; version 2.1; www. worldclim.org, Fick & Hijmans, 2017) using site location information (latitude and longitude). Future monthly values were estimated from the NCAR Community Model version 3 ( $2 \times CO_2$  climate change



FIGURE 1 Map of the 34 sites of study located within Europe and over the mean annual temperature (a) and the annual precipitation (b) gradients

## Mean Annual Temperature

ABLE 1 Characteristi	ics of samplin	ig sites and cov	/er (%) of each	ן plant species st	udied							4
Basin name	Latitude	Longitude	Altitude (m a.s.l.)	Lithology	Land type	Emberger's coefficient (Q2 )	Climate category according to Q2	Alnus glutinosa	Salix atrocinerea	Nerium oleander	Rubus ulmifolius	-WH
Rambla de las Negras	36.89	-2.01	47	Calcareous	Lowland	31.5	Semi-arid			13.8		LE
Río Aguas	37.09	-2.07	270	Calcareous	Lowland	33.8	Semi-arid			2.0	1.5	Y-
Barranco del Cura	36.84	-2.64	291	Calcareous	Lowland	40.4	Semi-arid			22.3	1.0	Div
Arroyo de Aguamulas	38.05	-2.82	689	Calcareous	Midland	42.6	Semi-arid				8.8	ersi
Río Bacares	37.31	-2.44	943	Calcareous	Midland	51.9	Sub-humid		22.8		25.7	<b>ty</b> ar
Río Vacal	36.92	-3.81	956	Calcareous	Midland	54.5	Sub-humid				6.8	nd Di
Arroyo Los Marcos	37.30	-2.58	1019	Calcareous	Highland	55.1	Sub-humid		20.5		10.3	istri
Río Mecina	36.99	-3.15	1136	Siliceous	Highland	56.5	Sub-humid	11.0	24.8		9.2	buti
Río Turrillas	36.93	-3.78	991	Calcareous	Midland	59.0	Sub-humid				19.3	ons
Río Andarax	37.01	-2.89	1013	Siliceous	Highland	60.4	Sub-humid	43.8	12.8			
Río Alhama	36.95	-3.96	959	Calcareous	Midland	61.2	Sub-humid				7.3	
Ribera de Santa Ana	37.87	-6.70	546	Siliceous	Midland	62.5	Sub-humid	66.3			9.3	
Arroyo Corterrangel	37.94	-6.60	462	Siliceous	Lowland	62.6	Sub-humid	15.8			12.7	
Río Nacimiento	37.15	-2.91	1149	Siliceous	Highland	63.6	Sub-humid		17.9		2.4	
Barranco del Dun Dun	37.94	-6.64	554	Siliceous	Midland	66.3	Sub-humid	52.5		1.5	3.0	
Río Guadalentin	37.89	-2.85	1273	Calcareous	Highland	66.3	Sub-humid		22.9		5.5	
Arroyo de Aguascebas	38.10	-2.87	1063	Calcareous	Highland	66.9	Sub-humid		3.0		4.0	
Arroyo de la Garganta	37.90	-2.89	1356	Calcareous	Highland	67.1	Sub-humid				11.6	
Río Chico Ohanes	37.05	-2.76	1038	Siliceous	Highland	72.1	Sub-humid		34.8	1.5	20.1	
Barranco del Pueblo	37.15	-3.15	1394	Siliceous	Highland	74.8	Sub-humid	29.7	6.0		3.5	
Arroyo Hondo	37.14	-3.03	1438	Siliceous	Highland	75.3	Sub-humid	51.4	3.5		15.0	
Arroyo de los Castaños	37.15	-3.01	1321	Siliceous	Highland	75.3	Sub-humid	62.0	16.3		2.8	
Arroyo de los Caballos	36.68	-4.91	350	Calcareous	Lowland	77.5	Sub-humid			17.7	7.1	
Arroyo de los Molinos	36.81	-5.37	379	Calcareous	Lowland	85.0	Sub-humid			11.3	8.2	
Arroyo de Bocaleones	36.83	-5.41	316	Calcareous	Lowland	88.6	Sub-humid			11.3	9.5	
Arroyo Gaidovar	36.78	-5.37	695	Calcareous	Midland	92.3	Humid				36.8	
Río Verde	36.66	-5.01	662	Calcareous	Midland	94.5	Humid			32.4	10.0	
Arroyo de la Cruz	36.64	-5.03	897	Calcareous	Midland	94.5	Humid				1.2	
Garganta del Caballo	36.54	-5.64	401	Siliceous	Lowland	102.2	Humid			11.9	35.9	R
Garganta del Aljibe	36.54	-5.63	432	Siliceous	Lowland	102.2	Humid	35.9			1.3	UBI
Garganta del Medio	36.54	-5.64	423	Siliceous	Lowland	102.2	Humid	17.3			2.3	O-RÍ
Garganta de la Cierva	36.52	-5.64	550	Siliceous	Midland	102.2	Humid	42.9		6.7	8.5	OS e

ulmifolius Rubus 6.3 6.3 Notes: Land type was defined by altitude (m a.s.l.): Low <500, Mid 500-1000, High >1000. Sites ranked in order of decreasing aridity according to the Emberger's bioclimatic coefficient Q2 oleander Verium atrocinerea Salix glutinosa Alnus 30.5 28.0 **Climate category** according to Q2 Humid Humid coefficient (Q2) Emberger's 109.5 109.5 Lowland Midland Land type Lithology Siliceous Siliceous Altitude (m a.s.l.) 429 532 Longitude -5.53 -5.53 Latitude 36.12 36.11 Garganta del Tesorillo Garganta de la Garza **Basin name** 

TABLE 1 (Continued)

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scenario, CCM3) for the year 2100 (Govindasamy et al., 2003) and subsequently downscaled and matched to the WorldClim estimates of current climate at a resolution of 2.5 minutes (i.e.  $\sim$ 4.5 × 4.5 km). From these variables, the Emberger's bioclimatic coefficient (Q2) for each site was calculated following Condés and García-Robredo (2012) as 100P/(( $M^2-m^2$ )), where *P* is the annual rainfall in mm, *M* the average maximum of the warmest month, and *m* the average minimum of the coldest month. To measure soil variables, we collected a sample consisting of six core samples of the top 20 cm of the riverbank soil profile, obtained by a randomly stratified method from each stream side at a distance of approximately 3 m from the active channel. Samples from each site were mixed, air dried, sieved (2 mm) and stored in sealed polyethylene bags until analysed. Soil physical and chemical variables (Table S1) were measured as in Gil et al. (2004).

## 2.3 | Leaf traits

We measured nine leaf traits that often correlate with leaf litter decomposition rate (see Graca et al., 2015; Tonin et al., 2021) for each species: N, P, Ca, Mg, condensed tannins and lignin concentrations, C:N and C:P molar ratios and toughness. Before measurements, leaves were rehydrated by spraying with distilled water and stored for 12 h at 5°C. Leaf toughness, expressed in units of mass (g), was measured by performing distal and proximal punctures per individual leaf using a Texture Analyzer TA.XTPlus (Stable Micro Systems) equipped with a needle of 0.38 mm<sup>2</sup> tip surface. Thereafter, leaves were oven dried (60°C, 72 h) and ground to fine powder (Mixer Mill RETSCH MM 200). Concentrations of C and N (% dry mass, DM) of leaves were determined using a mass spectrometer (EA-Thermo DELTA V Advantage. Fisher Scientific<sup>®</sup>) following standard procedures (Flindt et al., 2020). The concentration of P (% DM) was measured spectrophotometrically after autoclave-assisted extraction (APHA 1998, Flindt et al., 2020). Concentrations of Ca and Mg (% DM) were determined by inductively coupled plasma mass spectrometry (ICP-MS, Perkin Elmer DRC II). Condensed tannins (mg Catechin Hydrate Equivalent per g of DM) were measured by the acid butanol assay (Gessner & Steiner, 2020). Concentration of lignin (% DM) was estimated gravimetrically using the acid detergent method of Goering and Van Soest (1970).

## 2.4 | Data analysis

To elucidate the relationships between species cover and environmental variables, we ran a Canonical Correspondence Analysis (CCA; *cca* function of the 'vegan' package, Oksanen et al., 2019), after a forward selection (*ordistep* function of the 'vegan' package with 9999 permutations) of the most parsimonious subset of explanatory variables (PPSeasonality, PWettestM, MaxT, MinT and soil pH). Significance of all testable fractions was assessed using permutation tests. Environmental variables were transformed to improve the structure of the residuals using *log* or *arcsin* transformations for decimal and percentage values, respectively. — Diversity and Distributions

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Differences in individual traits among species were assessed using one-way ANOVA and post hoc Tukey tests (*anova* and *TukeyHSD* functions of the 'stats' package). We performed Principal Component Analyses (PCAs; *prcomp* function in the 'stats' package) to examine patterns in leaf trait variability: one pooling the four species to examine interspecific variation vs. trait plasticity, and one for each species to extract the main gradients (2 first PCs) of trait plasticity (i.e. leaf



FIGURE 2 Present (open, 2000) and projected (closed, 2100) Emberger's bioclimatic coefficient values (Q2), estimated from the NCAR Community Model version 3 (CCM3) for the year 2100 (Govindasamy et al., 2003), for each of the 34 streams studied. Note that higher Q2 values denote lower aridity

quality). Previously, using Spearman rank correlations, leaf traits with high (>0.85) collinearity were removed (Figure S1). Seven traits were finally included in the PCA: N, P, Ca, Mg, condensed tannins, lignin and toughness. *Log* or *arcsin* transformations of variables were used when required in ANOVA and PCA analyses. The relative magnitude of interspecific variation vs. species plasticity for the overall pool of traits for each species was estimated as the proportion that each species covered in each of the dimensions of the general PCA. Besides, to quantify the relative magnitude of interspecific variation vs. species plasticity for each leaf trait, we performed variance partitioning analyses (*varcomp* function of the 'ape' package, Paradis & Schliep, 2019).

We carried out partial least squares regressions (PLS; plsr function in the 'pls' package, Mevik et al., 2020) to evaluate the relative importance of climate and soil as predictors of leaf trait plasticity (first two PCA axes). Preliminary PLS regressions for each environmental matrix and plant species (Table S2) were used to reduce the number of variables by selecting those with the highest variable importance in projection (VIP; VIP function in the 'plsVarSel' package, Mehmood et al., 2012). Those variables with VIP ≥1 were considered relevant (Andersen & Bro, 2010). Spearman rank correlation analyses were used to equalize the size of the two matrices of environmental variables removing those variables with high collinearity within those with higher VIP values (Figure S2, Tables S3 and S4). A second PLS regression was performed for each species using the selected variables, and the influence of each group of environmental variables (climate and soil) and their combination (climate + soil) on leaf plasticity was assessed using the goodness of prediction  $(Q^2)$ and the goodness of fit ( $R^{2}(Y)$ ) of models. A model was considered significant when  $Q^2 > 0.097$  (Friden et al., 1994).

TABLE 2 Summary of univariate dependent variable PLS models fitted to the first two principal components of PCA (PC1 and PC2), summarizing leaf trait plasticity for each species, using three matrices (C, S and C+S) of selected (in preliminary PLS regressions) environmental variables as predictors

			Dep	endent variable	e			
		Cat of any incompanyal	PC1			PC2		
Functional type	Plant species	predictors or combination	N	Q <sup>2</sup>	R <sup>2</sup> (Y)	N	Q <sup>2</sup>	R <sup>2</sup> (Y)
Deciduous N-fixer	Alnus glutinosa	Climate (C)	1	0.63	0.76	0	-	-
		Soil (S)	4	0.26	0.74	1	0.09	0.43
		C+S	1	0.64	0.78	1	-0.02	0.37
Deciduous	Salix atrocinerea	Climate (C)	2	0.51	0.74	1	0.10	0.42
		Soil (S)	1	0.23	0.47	2	0.71	0.91
		C+S	1	0.36	0.59	4	0.58	0.92
Evergreen	Nerium oleander	Climate (C)	3	0.30	0.71	1	-0.02	0.46
		Soil (S)	1	-0.01	0.42	2	0.15	0.60
		C+S	6	0.70	0.98	1	0.13	0.54
Semi-deciduous	Rubus ulmifolius	Climate (C)	1	0.32	0.41	1	-0.05	0.09
		Soil (S)	1	0.03	0.19	1	-0.02	0.11
		C+S	1	0.26	0.40	2	0.06	0.32

Notes: The number of PLS dimensions with lowest cross validation error (N), goodness of prediction ( $Q^2$ ) and coefficient of determination of dependent variable ( $R^2$ ) are shown for each model. Significant models ( $Q^2 > 0.097$ ) are in bold.

In PLS regressions, all explanatory variables were scaled to unit variance (scale function) to give all variables the same relative importance. Regressions were carried out separately for each species and the number of extracted components (latent variables) and the robustness of the resulting models were determined by leaveone-out cross-validation (LOO). For each model, we determined the number of dimensions with the lowest cross-validation error. PLS regressions built with climatic variables, when statistically significant ( $Q^2 > 0.097$ ; Table 2), were used to estimate the projected change of leaf quality under the forecasted climate change scenarios for 2100, using the predict function of the 'stats' package. Current and projected values of leaf quality (i.e. mean position over PC 1 or PC 2 of separate PCAs for each species) were compared using t-tests for paired samples. Hedge's g effect size was estimated using the cohen.d function of the 'effsize' package (Torchiano, 2020). See Supporting Methods in Supplementary Material for further details of data analyses.

## 3 | RESULTS

# 3.1 | Environmental variables and species distribution

Overall, the four species covered a large gradient of climatic conditions from semi-arid to humid bioclimatic types according to the Emberger's coefficient (Q2) (Table 1, Figure 2). The scenario for 2100 developed by the NCAR Community Model version 3 (CCM3) (Govindasamy et al., 2003) forecasts a significant aridification in the studied region, greater in presently humid sites (Figure 2).

Distribution of each species was rather clearly separated by the environmental gradients established by the first two dimensions of the CCA (p < .001; Figure 3a), which explained 91% of fitted constrained variation (53% of total variation). Soil pH was the most important variable explaining species distribution, given its significant and positive load (0.72) on CCA 1 (also loading - 0.60 on CCA 2). Moreover, minimum annual temperature was positively correlated (0.43) with CCA 1. The above variables essentially determined the separation of the acidophilic A. glutinosa (hereafter Alnus) from other species, particularly from N. oleander (hereafter Nerium), which can tolerate high pH soils and prefers lowland sites with mild winters (Figure 3a, b, d; Table S5). Precipitation seasonality and precipitation of the wettest month loaded significantly and positively (0.70 and 0.53, respectively) on CCA 2. This dimension basically segregated S. atrocinerea (hereafter Salix), abundant at low-precipitation and neutral to basic soil sites, from other species (Figure 3c; Table S5). Rubus ulmifolius (hereafter Rubus) showed its highest cover at sites with basic soils and/or mild winters, where deciduous tree species developing dense canopy cover (alder, willow or other) were absent or scarce (Figure 3e; Table S5). The studied species varied in the range of environmental conditions they occupied. Rubus was the most widely distributed species, occupying 100% and 92% of CCA1 and CCA2 gradients, respectively, followed by Nerium (80.1% of CCA1

and 84.2% of CCA2), *Alnus* (45.5% of CCA1 and 78.5% of CCA2) and *Salix* with the most constricted distribution (44.6% of CCA1 and 71.7% of CCA2) (Table 1, Tables S1 and S2; Figure 3).

# 3.2 | Interspecific variation and species plasticity of leaf traits

Species differed significantly in all leaf traits measured (one-way ANOVAs, all p < .0001) (Figure 4, Table S6). Alnus showed the lowest toughness and the highest N concentration, and consequently the lowest C:N ratio, being for these traits antithetical to *Nerium*, which in turn showed the highest Ca concentration and C:P ratio. *Salix* exhibited the highest P, tannins and lignin concentrations, and *Rubus* the highest Mg concentrations. The first two components of the PCA on leaf traits for the four pooled species explained 62.5% of the variation (Figure 5a): PC 1 represented a gradient of increasing nutrients (N and P) parallel to decreasing leaf toughness, segregating the deciduous (*Alnus* and *Salix*) and semi-deciduous (*Rubus*) species, from the evergreen *Nerium* with the highest toughness and lowest nutrient concentrations. Tannins and lignin heavily loaded (0.67 and 0.80, respectively) on positive PC 2, where *Salix* samples were clustered.

Overall, interspecific variation was higher than trait plasticity (Figure 5a). *Rubus*, the most widely distributed species, showed higher trait plasticity on PC 1, occupying 54% of this leaf quality gradient while other species ranged between 23% and 38%. However, the two species with more restricted distribution, *Alnus* and *Salix*, showed the highest trait plasticity on PC 2, occupying 66% and 51% of this leaf quality gradient, respectively, compared to the more widely distributed *Rubus* and *Nerium* (both 40%) (Figure 5a).

Regarding individual traits, variance partitioning analyses indicated, overall, higher interspecific variation than species plasticity in leaf traits (Figure S3). The highest interspecific variation (>80%) occurred in traits considered major determinants of litter decomposability-palatability—toughness, lignin, N and C:N—as expected dealing with species across different plant functional types. However, trait plasticity was higher than interspecific variation for P, Ca, Mg and C:P (ranging between 55% and 71%) and noticeably high for tannins (Figure S3).

Trait plasticity was described by the first two principal components of the PCA performed for each species (Figure 5b–e), which explained a considerable proportion of trait plasticity: ranging between 54% in *Nerium* and 73% in *Alnus*. The first principal component (PC 1) represented for all species a gradient of increasing leaf quality (Figure 5b–e; Table S7) positively related with decomposability and digestibility, owing to its high positive correlation with leaf N (0.57–0.87) or P (0.56–0.93) concentrations, but negative with tannins (–0.90–0.29), lignin (–0.73–0.08) or toughness (–0.82 to –0.32). However, dimension PC 2 did not exhibit a common trend across species (Figure 5b–e; Table S7). For *Alnus* and *Salix*, PC 2 was positively correlated with N, but negatively with tannins and Mg, respectively; this component covaried positively with Ca but



FIGURE 3 Ordination of sampling sites for the first two CCA axes (90.87% of fitted constrained variance explained) based on the environmental variables. In a, vectors represent the weight of each environmental variable; open circles denote sampling sites; and squares show the mean weighted position of each plant species over the environmental gradient. In b-e, coloured circles and their sizes represent the sampling site where each plant species was present and the percentage of canopy cover in each site, respectively; black circles represent sampling sites where each corresponding species did not appear. Marginal density plots show distribution of data for parsimonious CCA values of each plant species

negatively with tannins in Nerium, and negatively with Ca, Mg, lignin and toughness in Rubus.

#### 3.3 Relative importance of climate and soil factors, and best climatic predictors of leaf trait plasticity

CCA2 (37.9% of fitted constrained, 22.1% of

Univariate dependent variable PLS models indicated that leaf trait plasticity (PC 1) of the four species responded significantly and predominantly to climatic variables (Table 2). Adding soil factors to climate increased noticeably the goodness of prediction in Nerium, but produced a highly complex model with six latent variables. Models predicting leaf trait plasticity associated to PC 2 were only significant for Nerium and Salix, but especially for the latter, in which the set of soil variables significantly predicted a high proportion of variance of leaf trait plasticity, but the model including just the set of climate variables was still significant (Table 2).

Overall, climatic predictors with the highest influence (VIP close or >1) on leaf trait plasticity associated to PC 1 (Table 3) varied among species, although most notable differences arose between broad functional groups. Mean temperature of the wettest quarter (late winter-early spring) was an important predictor with negative effects on leaf quality for deciduous/semi-deciduous species. Conversely, maximum annual temperature was the main predictor with high positive effect on leaf quality for the evergreen Nerium. Temperature annual range was an important predictor of leaf quality (PC 1) for Nerium and Alnus, although with contrasting sign (negative and positive, respectively), highlighting the opposite response that species belonging to different plant functional types may have the same climatic variable. Moreover, precipitation variables (Table 3) did not have substantial effects on the evergreen Nerium, but were important predictors of leaf quality (PC 1) for deciduous/semideciduous species, with notable positive effects on Salix and Rubus, but slightly negative on Alnus. Leaf quality of Salix associated to PC 2 was primarily predicted by temperature annual range (positive effect) and winter temperature (negative effect), with precipitation variables (Table 3) being other important predictors with positive effects on leaf quality. Over this dimension, soil EC and P (with negative effects) and soil CaCO<sub>3</sub> (with positive effects) were important predictors on leaf quality of Salix.

#### Forecasted intraspecific changes in leaf 3.4 quality induced by climate change

Our modelling projections showed that the four plant species would respond differently to the forecasted scenario of aridification by the year 2100 (2  $\times$  CO<sub>2</sub> climate change scenario) in the studied region, although with remarkable congruence within broad functional groups in terms of response direction (Figure 6; Figure S4). For Alnus and Salix (PC 1), we observed weak evidence of overall variation in leaf quality (t = 1.523, p = .154; t = -2.071, p = .065, respectively; Hedge's g = 0.232 and -0.295, respectively; Figure 6). Salix (PC 2;



FIGURE 4 Box-and-whisker plots for selected leaf trait variables of the four plant species studied: nitrogen (N), phosphorus (P), calcium (Ca) and magnesium (Mg) concentrations (% DM), molar elemental ratios (C:N and C:P), lignin concentrations (% DM), condensed tannins concentrations (mg Catechin Hydrate Equivalent g  $DM^{-1}$ ) and toughness (g) of each plant species. Box represents median and 25th and 75th percentile levels, crosses are the mean, whiskers are the range, and dots are replicates. Different letters indicate significant differences (p < .05) among plant species, on the basis of linear models followed by pairwise multiple comparisons (Tukey test)

Figure S4) and Rubus (PC 1; Figure 6) showed large (62% and 57% decrease in mean position, respectively) and significant depletion of their leaf quality (t = 2.423, p = .036; t = -8.277, p < .0001; respectively), with large effect size (Hedge's g = 1.242 and 1.064, respectively). Conversely, leaf quality of Nerium is projected to increase consistently and significantly (87% increase in PC 1; t = -8.277, p < .0001, Hedge's g = -1.437) in the scenario of rising aridity used for our predictions.

#### DISCUSSION 4

Functional trait-based approaches are potentially useful to understand how species respond to environmental changes (Soudzilovskaia et al., 2013; Zhang et al., 2020) and, therefore, are important for an ecologically sensitive management of ecosystems. Here, we assessed how climate change might affect leaf quality of different riparian woody species from an intraspecific perspective, which has been much disregarded based on the general assumption that intraspecific variation accounts only for an irrelevant portion of total trait variability (Garnier et al., 2001). Overall, in support of our first hypothesis, but contrary to previous studies (Graça & Poquet, 2014; Ordoñez et al., 2009), climate showed larger influence than soil explaining most leaf trait plasticity. Our second hypothesis of decreasing intraspecific leaf quality-linked to determinant traits of palatability and decomposability-with increasing aridity was partially supported, given that increasing temperature had negative effects on leaf quality of deciduous and semi-deciduous species, but not on the evergreen Nerium, which displayed the opposite response. These results suggest potential effects on stream ecosystem functioning (Fenoy et al., 2021; Martínez et al., 2013), but with inverse sign depending on the identity of dominant species in the riparian vegetation. Moreover, in support of our third hypothesis, we generally observed higher variation among species than plasticity within



FIGURE 5 Projection of the first two principal components of PCAs showing ordination of leaf samples of four species (circles) as a function of selected leaf traits (vectors). (a) interspecific variation (polygons of different colours) versus trait plasticity (circles in each polygon); (b–e) trait plasticity of each plant species separately: (b) Alnus glutinosa, (c) Salix atrocinerea, (d) Nerium oleander, (e) Rubus ulmifolius

species, except for a few traits (e.g. P, Ca and Mg concentrations and C:P ratio) that exhibited remarkable leaf trait plasticity (Albert et al., 2010; Fajardo & Piper, 2011). Nonetheless, ranges of trait plasticity found here for some traits (e.g. %N, %P and %lignin) are similar, or higher, than those reported before for other species (e.g. Lecerf & Chauvet, 2008; LeRoy et al., 2007; Oliveira et al., 2021).

## 4.1 | Relative influence of climate and soil factors

Over the environmental gradient studied, climate exhibited an overall higher influence than soil on most species' leaf trait plasticity, although soil was the strongest predictor in some cases (e.g. Salix and Nerium PC 2). We presumed higher responsiveness of leaf traits to climate than soil in species with distributions highly constrained by soil conditions. This appears to be the case for the acidophilic Alnus (Miles, 1985), the species with the highest control of climate on its leaf trait plasticity. The fact that Alnus is an N-fixer may have further contributed to make this species less sensitive to soil nutrients. However, other species with less restricted soil-related distributions, such as Rubus-spread out across almost the entire study area-or Nerium, also showed a prominent role of climate influence on leaf trait plasticity. Similar patterns have been observed when assessing the abundance of plant functional types in the same region and across environmental gradients (Salinas et al., 2018). This lower predictive role of soil variables may stem from the high dynamics of alluvial soils and their permanent water availability, which would tend to homogenize conditions-i.e. nutrient availability-among sites (Naiman & Decamps, 1997). Yet our results are counter to

other findings recorded at much larger spatial scales that observed substantial importance of soil predictors explaining intraspecific changes in leaf traits (Graça & Poquet, 2014; Ordoñez et al., 2009). This suggests that other factors not considered here, such as the great topographic variability present in our spatial gradient, or genotype differences, might be overriding soil effects.

## 4.2 | Main climatic predictors of leaf trait plasticity

Among climatic the factors, temperature exhibited much clearer patterns than precipitation on the main dimension of leaf trait plasticity (PC 1). This is to be expected in riparian belts of permanent streams where soil moisture tends to be relatively high and constant in the absence of extreme drought events (Moore et al., 2016), preventing major water stress in plants and its consequences on leaf characteristics (e.g. García-Palacios et al., 2016; LeRoy et al., 2014). However, climate-driven changes in streamflow may worsen the effects of aridification on such ecosystems (Perry et al., 2012).

Despite clear differentiation in distribution extent among species, we detected a common negative relationship between temperature and leaf quality in the deciduous and semi-deciduous species. On the contrary, this relationship was positive for the evergreen *Nerium*. Overall, nutrient concentrations (N, P, Ca and Mg) decreased, but tannin and/or lignin concentrations, and/or toughness increased with increasing temperature for deciduous/semi-deciduous species, whereas *Nerium* roughly exhibited the opposite pattern. Thus, within the frame of the leaf economic spectrum (Reich et al., 1997; Wright et al., 2004), the above seems to reveal antithetical syndromes of

Pct         Pct <th></th> <th>Alnus gl</th> <th>utinosa</th> <th>Salix at</th> <th>rocinerea</th> <th></th> <th></th> <th>Nerium</th> <th>oleander</th> <th></th> <th></th> <th>Rubus ulmi</th> <th>olius</th>		Alnus gl	utinosa	Salix at	rocinerea			Nerium	oleander			Rubus ulmi	olius	
Image: contractione with the		PC1		PC1		PC2		PC1		PC2		PC1		
Indefinition         0.03         0.03         0.03           Teasonity         1.05         0.06         1.1         0.05           Mart         1.1         0.05         1.11         0.05           Mart         1.3         0.05         1.03         0.04           Mart         1.3         0.05         1.03         0.04           Mart         0.05         0.05         1.03         0.04         0.36           Temus/Range         1.03         0.05         0.04         0.36         0.36           Temus/Range         0.05         0.05         0.04         0.36         0.36           Temus/Range         0.05         0.04         0.36         0.36         0.36           Temus/Range         0.05         0.04         0.36         0.36         0.36           Temus/Range         0.31         0.31         0.37         0.35         0.33           Temus/Range         0.31         0.31         0.36         0.33         0.33           Temus/Range         0.31         0.31         0.33         0.33         0.33           Temus/Range         0.31         0.31         0.33         0.33         0.33 <th>Climatic predictor</th> <th>٩١</th> <th>Standardized coefficient</th> <th>٨IP</th> <th>Standardized coefficient</th> <th>٩IJ</th> <th>Standardized coefficient</th> <th>٨I</th> <th>Standardized coefficient</th> <th>VIP</th> <th>Standardized coefficient</th> <th>VIP</th> <th>Standardized coefficient</th>	Climatic predictor	٩١	Standardized coefficient	٨IP	Standardized coefficient	٩IJ	Standardized coefficient	٨I	Standardized coefficient	VIP	Standardized coefficient	VIP	Standardized coefficient	
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Marf         1.13         0.54           TamaRange         103         0.06         1.04         0.36           Toularange         103         0.05         1.03         0.05         1.04         0.36           Toularange         0.95         0.05         1.33         0.021         1.33         0.021         1.00           Toularange         0.95         0.05         1.33         0.021         0.03         0.03           Toularange         0.95         0.05         0.13         0.021         0.03         0.04           Toularange         0.97         0.05         0.13         0.021         1.03         0.03           Toularange         0.97         0.05         0.14         0.14         0.14         1.09           Polescondity         0.97         0.04         0.04         0.04         0.04         1.00           Polescondity         0.97         0.04         0.04         0.04         0.04         1.00           Polescondity         1.14         0.14         0.14         1.10         1.10         1.10           Polescondity         1.14         0.14         1.14         1.14         1.14         1.10	TSeasonility	1.05	0.06											
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Interaction           Interaction           EC         1.15         -0.21           PH         -0.2           PH         -0.2           CaC03         -0.16         -0.03           CaC03         -0.16         -0.05         -0.04         -0.05           CaC03         -0.05         -0.16         -0.12         -0.05 <th cols<="" td=""><td>PWarmestQ</td><td></td><td></td><td></td><td></td><td>0.95</td><td>0.04</td><td></td><td></td><td></td><td></td><td></td><td></td></th>	<td>PWarmestQ</td> <td></td> <td></td> <td></td> <td></td> <td>0.95</td> <td>0.04</td> <td></td> <td></td> <td></td> <td></td> <td></td> <td></td>	PWarmestQ					0.95	0.04						
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CaCO <sub>3</sub> 0.93         -0.07         1.16         0.12         -0.01	Hd			1.15	-0.08					0.60	0.03			
Organic       0.71       -0.01         CEC       1.00       -0.15       1.28       -0.15         BasSat       0.80       -0.05       0.99       -0.07       -0.09         ESP       0.64       -0.02       0.59       -0.11       1.23       -0.15         P       1.40       0.27       0.90       0.06       1.11       -0.1       1.23       0.16	CaCO <sub>3</sub>			0.93	-0.07	1.16	0.12							
CEC         1.00         -0.15         1.28         -0.15           BasSat         0.80         -0.05         0.99         -0.07         0.7         -0.09           ESP         0.64         -0.02         0.59         -0.1         -0.15         -0.15           P         1.40         0.27         0.90         0.06         1.11         -0.1         1.23         0.16	Organic C									0.71	-0.01			
BasSat         0.80         -0.05         0.99         -0.07         0.7         -0.09           ESP         0.64         -0.02         0.59         -0.1         1.10         0.27         0.90         0.06         1.11         -0.1         1.23         0.16	CEC	1.00	-0.15							1.28	-0.15			
ESP         0.64         -0.02         0.59         -0.1           P         1.40         0.27         0.90         0.06         1.11         -0.1         1.23         0.16	BasSat	0.80	-0.05	0.99	-0.07	0.7	-0.09							
P 1.40 0.27 0.90 0.06 1.11 -0.1 1.23 0.16	ESP	0.64	-0.02			0.59	-0.1							
	Ь	1.40	0.27	0.90	0.06	1.11	-0.1			1.23	0.16			

TABLE 3 Variable importance in the projection (VIP) and standardized coefficients of the environmental variables used in the PLS models developed for the first dimension of the PCA (PC1)

Precipitation Seasonality (%); PWettestQ, Precipitation of Wettest Quarter (mm); PDriestQ, Precipitation of Driest Quarter (mm); PWarmestQ, Precipitation of Warmest Quarter (mm); EC, Soil Electric conductivity (μS cm<sup>-1</sup>); pH, Soil pH; CaCO<sub>3</sub>, Soil CaCO<sub>3</sub> (%); Organic C, Soil Organic Carbon (%); CEC, Soil Cation exchange capacity (cmol<sup>+</sup> kg<sup>-1</sup>); BasSat, Soil Base saturation (%), ESP, Soil Exchangeable Sodium Percentage (%); P, Soils Phosphorus (%). Relevant variables (VIP > 1) are in bold.

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FIGURE 6 Boxplots showing (a) the aridity range covered by the distribution of each species according to the Emberger's bioclimatic coefficient (Q2); and (b) the leaf quality (PC1) of each plant species at present (Current) and in future climate change scenarios according to the NCAR Community Model version 3 (CCM3) for the year 2100 (Projected). Note that leaf quality ranges (PCA dimensions) are scaled to unit for simplicity. Different letters indicate significant different based on t-test analyses

leaf traits between functional groups in response to temperature, in which the intraspecific intercorrelated leaf traits along our quality gradient represent physiological and structural trade-offs (Boyero et al., 2017; Onoda et al., 2017).

Decreasing leaf N and/or P concentrations with increasing temperature has been reported before in woody deciduous species (Chen et al., 2011; Kudo et al., 2001; Sun et al., 2015). This may be explained by an increase of the catalytic capacity of photosynthetic enzymes at higher temperatures, requiring lower enzyme amounts (e.g. lower N concentration) to maintain photosynthetic rates (i.e. the photosynthetic rate is achieved with lower amounts of such enzyme; Scafaro et al., 2017). Alternatively, or additionally, higher temperature is often associated with increasing length of the growing season in deciduous species, which in turn promotes long leaf life span. Long-lived leaves often invest more in structure/protection at the expense of reducing photosynthetic efficiency (Kudo et al., 2001; Onoda et al., 2017). Similarly, other authors have reported that deciduous plants growing under relatively elevated temperatures develop

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tougher leaves (Wright et al., 2017) or leaves with higher tannin (Top et al., 2017) and lignin (Graça & Poquet, 2014) concentrations.

Reduction of nutrients and strengthening of leaf traits to confer resistance (e.g. increasing toughness) have been reported in evergreen Quercus species in response to decreasing winter temperatures. This is interpreted as a higher cost for evergreens at cooler sites compared with deciduous trees (González-Zurdo et al., 2016). However, this finding is not totally consistent with our results for Nerium as winter temperatures did not exhibit any effect on its leaf quality. We observed the strongest positive effect on leaf quality of Nerium from maximum temperature, but a more negative effect from annual temperature range. This suggests that Nerium develops more nutrient-rich and softer leaves in its optimum distributional range (areas with mild winters and maritime influence), with negligible effects from harsh low-winter temperatures, which are infrequent in its area of distribution. Nevertheless, we cannot rule out the possibility that our results are species-specific, and projection of such results to the entire functional group needs to be confirmed with the study of further evergreen species.

A substantial amount of leaf trait plasticity (25%) in *Salix* (PC 2)—positively related to leaf N and lignin, and negatively to Mg concentrations—was significantly explained by climatic conditions, but much more by soil variables. The strong positive association of N and lignin on PC 2 suggest that this N fraction is structural, possibly lignin-bound N, therefore not readily available to decomposers and detritivores (Berendse et al., 1987). Thus, PC 2 represents a structural reinforcement of *Salix* leaves positively related with temperature annual range and negatively with winter temperature, but also, and mostly, negatively with soil P. A structural reinforcement of leaves (increasing leaf mass per area and lignin concentration) with decreasing soil fertility has been documented elsewhere (e.g. Diehl et al., 2008).

The trait plasticity observed in this study can arise from responses to environmental conditions, but also from genetic variability. Genotypes, although largely influenced and selected by local environments, represent an important source of trait variability unaccounted for here. Genetic variability has been exhibited to strongly influence litter quality and, consequently, associated ecosystem processes (e.g. litter decomposition) and communities (Crutsinger et al., 2014; LeRoy et al., 2006, 2007, 2012). Given that leaf traits differ in their heritability, for example, tannins appear to be highly heritable whereas C:N ratios are environmentally controlled (Crutsinger et al., 2014), further research assessing how environment × genotype interaction affects leaf traits is important for improving predictions of potential effects on ecosystem functioning, particularly, in the face of climate change.

## 4.3 | Projecting climate change-driven variation of species leaf quality: implications for stream ecosystems

Litter trait variation across species constitutes the main driver of instream litter decomposition worldwide (Boyero et al., 2017;

García-Palacios et al., 2016; Zhang et al., 2019), indicating an essential role of plant phylogenetic history on controlling such process (LeRoy et al., 2019). Although less studied, some evidence indicates that the control exerted by trait plasticity on litter decomposition, nutrient cycling and trophic dynamics could be almost as important as interspecific changes (Jackrel & Morton, 2018; Jackrel et al., 2016; Lecerf & Chauvet, 2008; LeRoy et al., 2007; Oliveira et al., 2021). Here, we assessed the plasticity of selected traits of green leaves of riparian plants aimed at forecasting potential consequences of climate change on stream ecosystems highly dependent on these resources (i.e. forest streams; Wallace et al., 2015). Although inputs of leaves to streams are mainly in the form of leaf litter, it has been reported that some traits of green leaves tend to persist after senescence and control rates of litter decomposition (Cornelissen et al., 1999; Cornwell et al., 2008). Therefore, if nutrient resorption efficiency remains fundamentally invariable across climatic conditions (Norby et al., 2000, Aerts et al., 2007, but see Yuan & Chen, 2009b), understanding how green leaves respond to climate change may allow us to anticipate effects of leaf quality changes on stream ecosystem functioning. In support of this idea, a recent study suggests that traits of green leaves can be used to accurately predict decomposition rates (Rosenfield et al., 2020). However, as others have pointed out that traits of litter can differ from those of fresh leaves (Hättenschwiler et al., 2008; Hättenschwiler & Vitousek, 2000; Horner et al., 1987; Yuan & Chen, 2009a), the potential effects on headwater stream functioning exposed here should be interpreted with caution.

Litter decomposition is often reported to be enhanced by its high N and P concentrations (García-Palacios, McKie, et al., 2016; MacKenzie et al., 2013). Elevated litter concentrations of Ca and Mg-reported to be important for fungal decomposers (Jenkins & Suberkropp, 1995) and macroinvertebrates (Makkonen et al., 2012; National Research Council, 2005)—can also accelerate decomposition (Santonja et al., 2019). Moreover, tannins (Coq et al., 2010; Irons et al., 1988), lignin (Ferreira et al., 2016; Ramos et al., 2021; Schindler & Gessner, 2009) and toughness (Fenoy et al., 2021; Li et al., 2009) primarily tend to reduce litter consumption by detritivores. Our results point to a general decrease in leaf quality as a response to aridification in the three deciduous/semi-deciduous species. This decrease was generally related to a reduction in leaf N and P, but also Ca and Mg, versus an increase in tannins or lignin, and leaf toughness.

In particular, changes in leaf quality of the deciduous N-fixer Alnus could have major consequences given the key role of this species on stream ecosystem processes (Alonso et al., 2021; Pérez, Basaguren, et al., 2021; Rubio-Ríos et al., 2021). We reported here for Alnus ranges of %N, %P and %lignin variation similar to those reported at the European continental scale (Lecerf & Chauvet, 2008), and 53% of its species leaf trait plasticity was remarkably explained by climatic variables, yet our forecasted decrease in leaf quality was relatively low (11%) and not statistically significant, compared to other species. Nonetheless, apparent subtle changes in litter traits 14

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might result in major effects in consumer fitness (Pérez et al., 2021). Furthermore, this projected minor decrease in leaf quality adds to the decline of populations of this key species through Europe due to a disease caused by the pathogen *Phytophthora alni* (Bjelke et al., 2016), which also has been recently reported to alter the nutritional quality of leaf litter (Ferreira et al., 2021). Both factors are likely to trigger significant alterations to the functioning of forested streams (Alonso et al., 2021). Moreover, if a general decrease in leaf quality occurs in other deciduous species, as those forecasted here for *Salix* and *Rubus*, the negative influences on stream food webs will increase.

Thus, our results indicated that decreases of leaf quality of individual deciduous species may occur in a relatively short term (via phenotypic plasticity; Nicotra et al., 2010; but see Valladares et al., 2007), which in the long term will add to the forecasted dieback of deciduous woody species in riparian corridors of temperate climate zones (Kominoski et al., 2013; Salinas et al., 2018). Both riparian changes have the potential to significantly impair instream ecosystem processes, particularly in mountain streams presently dominated by deciduous vegetation (Fenoy et al., 2021), more than in lowland streams where deciduous species actually represent a minor component of the riparian belt.

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#### CONFLICT OF INTEREST

The authors declare no conflict of interest.

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#### DATA AVAILABILITY STATEMENT

Data openly available in a public repository https://doi.org/10.5061/ dryad.bzkh1899h.

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## BIOSKETCH

The research team is made up of ecologists and botanists whose work focuses on the basic and applied aspects of the ecology of streams and riparian zones. We aim to increase our knowledge about the relationships between riparian plant communities and stream ecosystem functioning.

Author contributions: JJC and MJS connceptualized the study. JRR, MJS and JJC contributed to methodology. JRR performed analyses with help of JP and EF. JRR wrote the original draft with help of JP and JJC. All authors reviewed and edited the manuscript. JJC provided resources and performed funding acquisition.

#### SUPPORTING INFORMATION

Additional supporting information may be found in the online version of the article at the publisher's website.

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