Received: 17 September 2020 | Revised: 22 February 2021 | Accepted: 8 March 2021

DOI: 10.1111/geb.13297

RESEARCH PAPER

Global Ecology and Biogeography | A Journal of Macroecology | WILEY

Global patterns and drivers of alpine plant species richness

Riccardo Testolin[1] | Fabio Attorre[1] | Peter Borchardt[2] | Robert F. Brand[3,4] | Helge Bruelheide[5,6] | Milan Chytrý[7] | Michele De Sanctis[1] | Jiri Dolezal[8,9] | Manfred Finckh[10] | Sylvia Haider[5,6] | Andreas Hemp[11] | Ute Jandt[5,6] | Michael Kessler[12] | Andrey Yu Korolyuk[13] | Jonathan Lenoir[14] | Natalia Makunina[13] | George P. Malanson[15] | Daniel B. Montesinos-Tubée[16,17,18] | Jalil Noroozi[19] | Arkadiusz Nowak[20,21] | Robert K. Peet[22] | Gwendolyn Peyre[23] | Francesco Maria Sabatini[6,5] | Jozef Šibík[24] | Petr Sklenář[25] | Steven P. Sylvester[12] | Kiril Vassilev[26] | Risto Virtanen[6,27,28] | Wolfgang Willner[19] | Susan K. Wiser[29] | Evgeny G. Zibzeev[13] | Borja Jiménez-Alfaro[30]

[1]Department of Environmental Biology, Sapienza University of Rome, Rome, Italy

[2]Institute of Geography, CEN – Center for Earth System Research and Sustainability, University of Hamburg, Hamburg, Germany

[3]Cuyahoga County Board of Health, Parma, Ohio, USA

[4]Department of Soil, Crop and Climate Science, University of the Free State, Bloemfontein, South Africa

[5]Institute of Biology/Geobotany and Botanical Garden, Martin Luther University Halle-Wittenberg, Halle, Germany

[6]German Centre for Integrative Biodiversity Research (iDiv) Halle-Jena-Leipzig, Leipzig, Germany

[7]Department of Botany and Zoology, Faculty of Science, Masaryk University, Brno, Czech Republic

[8]Institute of Botany of the Czech Academy of Sciences, Průhonice, Czech Republic

[9]Department of Botany, Faculty of Science, University of South Bohemia, České Budějovice, Czech Republic

[10]Biodiversity, Evolution and Ecology of Plants, Institute of Plant Science and Microbiology, University of Hamburg, Hamburg, Germany

[11]Department of Plant Systematics, University of Bayreuth, Bayreuth, Germany

[12]Department of Systematic and Evolutionary Botany, University of Zurich, Zurich, Switzerland

[13]Central Siberian Botanical Garden, Siberian Branch, Russian Academy of Sciences, Novosibirsk, Russia

[14]UR "Ecologie et Dynamique des Systèmes Anthropisés" (EDYSAN, UMR 7058 CNRS-UPJV), Université de Picardie Jules Verne, Amiens, France

[15]Department of Geographical & Sustainability Sciences, University of Iowa, Iowa City, Iowa, USA

[16]Naturalis Biodiversity Center, Leiden, The Netherlands

[17]Instituto Científico Michael Owen Dillon, Cercado, Arequipa, Perú

[18]Escuela Profesional de Ingeniería Ambiental, Universidad Nacional de Moquegua, Pacocha, Moquegua, Peru

[19]Department of Botany and Biodiversity Research, University of Vienna, Vienna, Austria

[20]Botanical Garden – Center for Biological Diversity Conservation, Polish Academy of Sciences, Warszawa, Poland

[21]Institute of Biology, University of Opole, Opole, Poland

[22]Department of Biology, University of North Carolina, Chapel Hill, North Carolina, USA

[23]Department of Civil and Environmental Engineering, University of the Andes, Bogota, Colombia

[24]Institute of Botany, Plant Science and Biodiversity Centre, Slovak Academy of Sciences, Bratislava, Slovakia

[25]Department of Botany, Charles University, Prague, Czech Republic

[26]Department of Plant and Fungal Diversity and Resources, Institute of Biodiversity and Ecosystem Research, Bulgarian Academy of Sciences, Sofia, Bulgaria

[27]Department of Physiological Diversity, Helmholtz Center for Environmental Research – UFZ, Leipzig, Germany

[28]Department of Ecology and Genetics, University of Oulu, Oulu, Finland

[29]Manaaki Whenua – Landcare Research, Lincoln, New Zealand
[30]Research Unit of Biodiversity (CSUC/UO/PA), University of Oviedo, Mieres, Spain

Correspondence
Riccardo Testolin, Department of Environmental Biology, Sapienza University of Rome, P.le Aldo Moro 5, 00185 Rome (RM), Italy.
Email: riccardo.testolin@gmail.com

Funding information
Leverhulme Trust; MSMT Inter-Excellence Project, Grant/Award Number: LTAUSA18007; Czech Science Foundation, Grant/Award Number: 19-28491X; US National Science Foundation, Grant/Award Number: 1853665; Schweizerischer Nationalfonds zur Förderung der Wissenschaftlichen Forschung, Grant/Award Number: P2ZHP3_161988; Principality of Asturias-EU, Grant/Award Number: ACB17-26; Deutsche Forschungsgemeinschaft, Grant/Award Number: DFG FZT 118, HE 6726 5/1, JU 2748/6-1 and 202548816; NZ Ministry for Business, Innovation and Employment's Strategic Science Investment Fund; Universität Zürich; Spanish Research Agency, Grant/Award Number: AEI/10.13039/501100011033

Editor: Anne Bjorkman

Abstract

Aim: Alpine ecosystems differ in area, macroenvironment and biogeographical history across the Earth, but the relationship between these factors and plant species richness is still unexplored. Here, we assess the global patterns of plant species richness in alpine ecosystems and their association with environmental, geographical and historical factors at regional and community scales.

Location: Global.

Time period: Data collected between 1923 and 2019.

Major taxa studied: Vascular plants.

Methods: We used a dataset representative of global alpine vegetation, consisting of 8,928 plots sampled within 26 ecoregions and six biogeographical realms, to estimate regional richness using sample-based rarefaction and extrapolation. Then, we evaluated latitudinal patterns of regional and community richness with generalized additive models. Using environmental, geographical and historical predictors from global raster layers, we modelled regional and community richness in a mixed-effect modelling framework.

Results: The latitudinal pattern of regional richness peaked around the equator and at mid-latitudes, in response to current and past alpine area, isolation and the variation in soil pH among regions. At the community level, species richness peaked at mid-latitudes of the Northern Hemisphere, despite a considerable within-region variation. Community richness was related to macroclimate and historical predictors, with strong effects of other spatially structured factors.

Main conclusions: In contrast to the well-known latitudinal diversity gradient, the alpine plant species richness of some temperate regions in Eurasia was comparable to that of hyperdiverse tropical ecosystems, such as the páramo. The species richness of these putative hotspot regions is explained mainly by the extent of alpine area and their glacial history, whereas community richness depends on local environmental factors. Our results highlight hotspots of species richness at mid-latitudes, indicating that the diversity of alpine plants is linked to regional idiosyncrasies and to the historical prevalence of alpine ecosystems, rather than current macroclimatic gradients.

KEYWORDS
Alpine vegetation, biodiversity hotspots, biogeographical history, global patterns, multiscale analysis, plant species richness

1 | INTRODUCTION

More than 200 years after the attempt by Alexander von Humboldt to formulate a unified theory of the natural world, understanding the global patterns of diversity remains one of the greatest challenges in biogeography and macroecology (Brummitt et al., 2020; Keil & Chase, 2019; Kier et al., 2005; Kreft & Jetz, 2007; Kreft et al., 2008; Weigelt et al., 2016). In particular, mountains have been revealed as centres of biodiversity, with a disproportionately high species richness in comparison to their corresponding lowland regions (Antonelli et al., 2018; Muellner-Riehl et al., 2019; Rahbek, Borregaard, Antonelli, et al., 2019). Along the elevational gradient of mountains, the compression of life zones brings different biomes into proximity, with the alpine belt representing the outpost for plant life above the climatic tree line. Alpine ecosystems, governed by low-temperature regimes, cover *c*. 3% of land

outside Antarctica and are distributed across all continents and latitudes (Körner et al., 2011; Testolin et al., 2020). Despite ongoing efforts to monitor changes in the biota of mountain summits in the face of climate change (Gottfried et al., 2012; Pauli et al., 2012; Steinbauer et al., 2018), we still lack a picture of the global patterns of plant diversity in alpine habitats, let alone an understanding of its major drivers.

According to the general latitudinal diversity gradient, biodiversity is expected to peak at the equator (Hillebrand, 2004). Among the possible explanations for this pattern (Lomolino et al., 2017), latitude is normally interpreted as a proxy for climatic conditions and available metabolic energy, which might have an effect on speciation rates (Wang et al., 2009). Whether this general rule also applies to alpine ecosystems, however, is still a matter of debate. Regardless of their latitude, alpine ecosystems are determined by low-temperature conditions, hence low energy input. Therefore, lowland and alpine thermal conditions from polar to equatorial latitudes are increasingly decoupled from one another (Testolin et al., 2020). Besides having a lower energy input compared with the lowlands, alpine ecosystems are also highly heterogeneous in their topoclimates (Quinn, 2008), which might weaken the correlation between latitude and primary productivity (Testolin et al., 2020). For these reasons, plant diversity in alpine areas might decouple from major climatic gradients.

Alpine areas are also isolated from each other, forming fragmented systems of "sky islands" surrounded by lowland environments that limit species dispersal (McCormack et al., 2009). Following the ecological principle of the species–area relationship (Lomolino, 2000b) and its application to the theory of island biogeography (MacArthur & Wilson, 1967), the extent of alpine habitats and their isolation could have affected rates of colonization, speciation and extinction of plants (Heaney, 2000; Steinbauer et al., 2016). These processes might have resulted in biodiversity patterns linked to the historical and current abundance of alpine habitats at the global scale. Although it has been reported widely that the biogeographical history of mountains has shaped diversity patterns of cold-adapted plant species in alpine regions (Flantua et al., 2019; Harris, 2007; McGlone et al., 2001; Sklenář et al., 2014), a major unresolved question is the extent to which the interplay of ecological drivers and historical contingencies dictates the patterns of alpine plant diversity at the global level (Nagy & Grabherr, 2009). The significance of these drivers might shift from global to local spatial scales and can reveal new patterns and relationships that are not evident at regional scales at which alpine plant diversity patterns have been studied so far (Jiménez-Alfaro et al., 2014; Lenoir et al., 2010; Moser et al., 2005; Vonlanthen et al., 2006).

Here, we compiled a dataset of 8,928 vegetation plots with 5,325 vascular plant species sampled by botanical experts in alpine ecosystems over the past 100 years and representative of global alpine vegetation. By analysing the data at both regional and community levels, we investigate: (a) the global latitudinal patterns of alpine plant species richness; and (b) the relative influence of environmental, geographical and historical factors in driving such patterns. We also evaluate how those patterns and drivers change between regional and community levels and how they relate to hotspots of alpine plant diversity recognized at the global scale.

2 | METHODS

2.1 | Study system and data collection

We considered as zonal alpine vegetation any plant community dominated by graminoids, forbs and dwarf shrubs above the climatic tree line (Körner, 2003). In addition to strictly zonal habitats, snow-patch vascular plant communities and vegetation on rocks and screes are also found ubiquitously in the alpine belt and were included in our study. We did not consider vegetation from polar climates owing to the absence of elevational tree lines and their distinct environments (Quinn, 2008; Walter & Box, 1976). Therefore, the alpine vegetation included in the present study corresponds to the "mid-latitude alpine tundra" and the "tropical alpine biome" groups as defined by Quinn (2008).

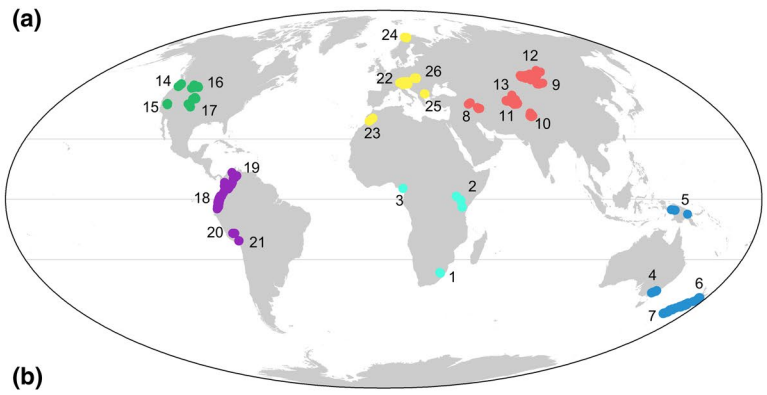
We assembled vegetation-plot data of vascular plant communities sampled by the authors, compiled from the literature or included in sPlot (the global vegetation database; Bruelheide et al., 2019), with the aim of obtaining a representative sample of the global alpine vegetation. The plots were selected using habitat classifications of the data sources and revised by data providers based on the scope of our study (Supporting Information Table S1). Datasets from different sources were standardized by identifying a minimum common set of plot attributes, including size, elevation and geographical coordinates. When the geographical coordinates were missing for small, clearly delimited areas, we estimated plot locations from maps (i.e., Mount Jaya; Hope et al., 1976) or by randomly assigning the coordinates of raster cells with the same elevation (±10 m) as the plots in that area (i.e., Mount Wilhelm and Drakensberg; Brand et al., 2015; Wade & McVean, 1969), using the SRTM-3 digital elevation model at 30 m resolution (Farr et al., 2007; NASA & JPL, 2013). Species cover values with discrete scales were transformed to the mean value of the corresponding percentage interval. Species names were harmonized using the Taxonomic Name Resolution Service (Boyle et al., 2013) online tool (https://tnrs.biendata.org/) with default settings, updating the names to the most recent nomenclature and merging subspecies and varieties to the species level by summing their respective cover values.

The initial dataset, consisting of 10,408 plots, was filtered further by removing plots with tree species or incomplete taxonomic identification. When taxa identified to the genus level or higher taxonomic rank represented ≥ 10% of the plot vegetation cover, the corresponding plot was discarded; otherwise, we removed those taxa from the plot record (3,086 plots from which at least one taxon was removed; median number of taxa removed = 1). Each plot was then assigned to a region based on its location. Regions were defined based on the approximate extent of ecoregions (Olson et al., 2001), which represent an ecologically meaningful framework for identifying distinct

geographical units at the global scale. Given that the names of some ecoregions did not reflect the presence of an alpine vegetation belt, we renamed these regions after the main mountain ranges where the plots were located, following Körner et al. (2017) (Supporting Information Table S2). For the analyses, we retained only regions with ≥ 60 plots and removed extremely small or large plots (<.25 or >400 m^2). To filter out compositional outliers, we performed a detrended correspondence analysis (DCA) on each regional dataset, excluding those plots whose score on the first axis (DCA1) was larger or smaller than 10 times the width of the interquartile range from the median. After removing the outliers, the gradient length of DCA1 ranged from 3.6 to 9.9 standard deviation units of species turnover within different regions (Supporting Information Table S3), indicating different, yet high, degrees of regional beta diversity. Finally, to assess the representativeness of our dataset, we compared the climatic space of the plots against the climatic envelope of global alpine areas (Testolin et al., 2020; Supporting Information Figure S1). The final dataset consisted of 8,928 plots of alpine vascular vegetation along elevational gradients above the tree line in 26 regions belonging to six biogeographical realms (Keil & Chase, 2019; Figure 1a,b; Supporting Information Table S3), distributed across all continents except Antarctica and sampled over a period of almost 100 years, between 1923 and 2019.

2.2 | Diversity measures

Given that the number of samples differed considerably among regions, we estimated regional species richness using sample-based rarefaction and extrapolation (Chao et al., 2014) with the software R (R Core Team, 2020) and the package iNEXT (Hsieh et al., 2016). This technique allows a statistically sound comparison of diversity across groups with different sample sizes through the construction of sampling curves for species richness. These curves can be rarefied (interpolated) to smaller sample sizes or extrapolated (predicted) to larger sample sizes (Chao et al., 2014; Hsieh et al., 2016). Here, we estimated the regional richness for a unique sample size of 180 plots, corresponding to approximately three times the smallest regional sample (Figure 1b,c). We chose 180 plots as a trade-off between the loss of data in intensively sampled regions versus the inclusion of all regions in the analyses. As such, these estimates should not be interpreted as representing the total regional species pools, but rather as comparable estimates of regional richness. Given that our global dataset comprised plots of different sizes, we evaluated the effect of plot size on the species richness estimates. To do this, we compared the same estimates using three subsets of different plot sizes (small, <10 m^2; medium, ≥ 10 and <100 m^2; and large, ≥ 100 m^2). For those



(b)

		Region	N	S_{est}		Region	N	S_{est}
Afrotropic	1	Drakensberg	91	212	Nearctic	14 Central and Southern Cascades	239	130
	2	Eastern African Mountains	238	176		15 Sierra Nevada	153	279
	3	Mount Cameroon	115	120		16 South Central Rocky Mountains	136	286
Australasia	4	Australian Alps	157	205		17 Southern Rocky Mountains	153	260
	5	Central Range New Guinea	91	157	Neotropic	18 Colombian and Ecuadorian Andes	1534	543
	6	North Island Mountains New Zealand	67	204		19 Cordillera de Mérida	74	350
	7	Southern Alps New Zealand	743	281		20 Southern Cordillera Occidental Peru	85	150
Eastern Palaearctic	8	Alborz Mountains	524	143		21 Southern Cordillera Oriental Peru	168	253
	9	Altai Mountains	544	486	Western Palaearctic	22 Central and Eastern Alps	684	387
	10	Ladakh Range	452	242		23 High Atlas Range	125	187
	11	Pamir Mountains	175	497		24 Northern Scandes	76	98
	12	Sayan Mountains	314	369		25 Rila	267	147
	13	Western Tien Shan	74	179		26 Western Carpathians	1649	252

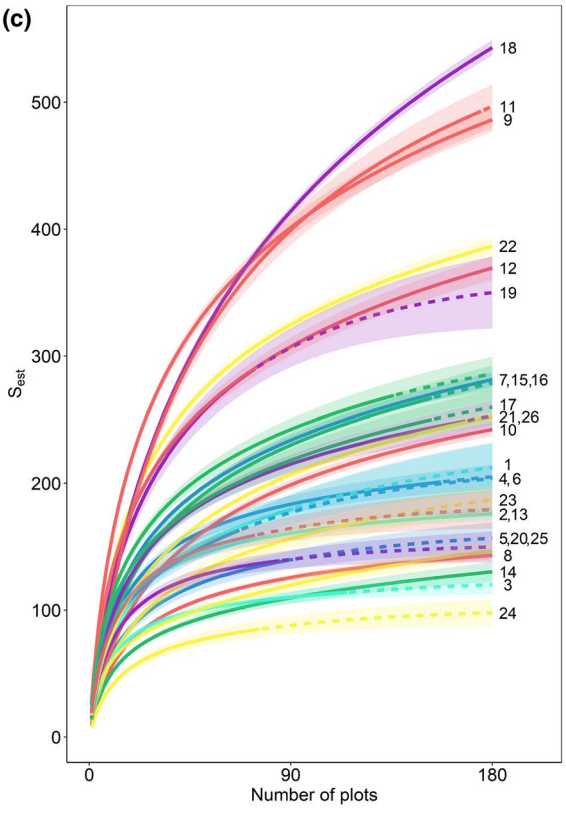


FIGURE 1 Overview of the alpine vegetation dataset and regional species richness. (a) Spatial distribution of alpine vegetation plots. (b) Number of plots collected in this study (N) and estimated species richness (S_{est}) for a comparable number of 180 plots in 26 alpine regions and six biogeographical realms. (c) Rarefaction curves of species richness for each region. Dashed lines indicate extrapolated values beyond the available number of plots. Continuous lines indicate that regional estimates were interpolated from larger sample sizes. The shaded areas represent the 95% bootstrap confidence intervals [Colour figure can be viewed at wileyonlinelibrary.com]

regions where ≥ 60 plots of each of the three different sizes were available (Alborz Mountains, Central and Eastern Alps, Colombian and Ecuadorian Andes, Eastern African Mountains, South Central Rocky Mountains and Western Carpathians), we compared richness estimates obtained from the different subsets. We found that, regardless of the subset used, the relative differences among regions were largely preserved, especially for those datasets comprising large numbers of plots (e.g., Central and Eastern Alps and Western Carpathians). An exception was the region of the Colombian and Ecuadorian Andes, where regional richness estimates were highly dependent on plot size (Supporting Information Figure S2). However, large and small plots both resulted in lower species richness estimates compared with medium-sized plots, suggesting that the differences are driven by different vegetation types being sampled with differently sized plots.

For each plot, we calculated community richness as the total number of species. We evaluated latitudinal patterns of regional and community richness using generalized additive models (GAMs), with a smoothing term for latitude. Our alpine regions were characterized by very different extents, and plot size varied widely. To account for different regional extents and plot sizes in the evaluation of species richness patterns, we also fitted GAMs on the residuals from ordinary least square regressions of ln(regional richness) as a function of ln(current local alpine area) and ln(community richness) as a function of ln(plot size). The procedure for calculating local alpine area is described in section 2.4.

2.3 | Environmental predictors

To analyse the drivers of species richness, we retrieved a set of environmental variables linked to plant diversity in the alpine belt from online sources. We calculated several climatic predictors at the plot level using digital sources at *c.* 1 km resolution. We used data from CHELSA (Karger et al., 2017) within the time frame of the growing season, defined as days with mean temperature > .9°C (Paulsen & Körner, 2014). Given that daily temperature data were not available, we estimated the growing season using monthly averages, including the months with a mean temperature > .9°C. Although this might have resulted in a sharper delimitation of season lengths, it probably had little effect on our global analyses. We included the mean temperature, precipitation, growing degree days and mean potential evapotranspiration of the growing season, all of which have been reported to have positive effects on photosynthetic activity and species richness in alpine areas (Körner, 2003; Moser et al., 2005; Nagy & Grabherr, 2009). Growing degree days (i.e., the sum of monthly temperatures > .9°C multiplied by the total number of days in such months) were calculated using the "growingDegDays" function of the R package envirem (Title & Bemmels, 2018). The mean potential evapotranspiration of the growing season was estimated with the "hargreaves" function of the R package SPEI (Beguería & Vicente-Serrano, 2017), using maximum and minimum monthly values of temperature and monthly precipitation. The monthly values of potential evapotranspiration obtained were then averaged across months with a mean temperature > .9°C.

Together with climate, soil pH is known to be a significant driver of species richness in the alpine belt (Vonlanthen et al., 2006) and is a good surrogate for the dominant bedrock, effectively differentiating calcareous and siliceous substrates (Jiménez-Alfaro et al., 2018; Lenoir et al., 2010). We derived estimates of soil pH from the SoilGrids database, averaging the values estimated at 5 and 15 cm depths (Hengl et al., 2017). When the pH value was missing for a plot location (45 plots), we assigned the value of the closest pixel to the plot. Despite the limitations posed by the use of global datasets to estimate fine-scale soil properties (Hengl et al., 2017), the obtained values covered a wide span of soil pH variation in alpine environments (4.40–8.35) and were, therefore, useful to distinguish dominant bedrock types.

In addition to climate and soil, topography also represents a major factor linked to plant diversity in alpine areas (Jiménez-Alfaro et al., 2014; Lenoir et al., 2010), because it creates a fine-scale mosaic of contrasting microclimates (Körner, 2003; Nagy & Grabherr, 2009; Rahbek, Borregaard, Antonelli, et al., 2019; Rahbek, Borregaard, Colwell, et al., 2019). As measures of topographical heterogeneity of the terrain surrounding each plot, we used the topographical position index and the terrain ruggedness index (Amatulli et al., 2018). Regional values of the predictors computed at the plot level were then estimated as the average of all vegetation plots within a region. For climatic predictors and soil pH, we also calculated the standard deviation of the predictor to test for the effect of environmental heterogeneity.

2.4 | Geographical and historical predictors

In addition to environmental variables, large-scale geographical factors, such as area and isolation, are known to influence the current diversity of island-like ecosystems (MacArthur & Wilson, 1967; Whittaker et al., 2008, 2017), as do their historical changes caused by climatic fluctuations (Fernández-Palacios et al., 2016; Weigelt et al., 2016). We delimited alpine area as the portion of land with a mean temperature of the growing season between 3.5 and 6.4°C or with a length of the growing season between 1 and 3 months (Paulsen & Körner, 2014). We did this both for current climatic conditions and considering climate during the Last Glacial Maximum (LGM) (Supporting Information Figure S3).

Alpine areas were calculated at two scales, reflecting the spatial extents of each regional sample (local area) and the total continuous alpine area extending beyond the samples (total area). We defined the local area as the extent of the alpine area contained within the convex hull formed by all plots of each region. In some cases, the coarse resolution of the climatic datasets used to estimate the alpine areas failed to detect any alpine patch within the hulls. Therefore, we applied a 5 km buffer around each hull to include at least some alpine area patches for all regions. The total alpine area for each region was estimated as the continuous extent of all alpine patches intersected by the hulls, reflecting the total extent of alpine habitats available

to species dispersal (Supporting Information Figure S4). Calculating alpine areas at two scales also allowed us to estimate the completeness of the regional samples by calculating the percentage of the local alpine area encompassed by the samples over the total alpine area (Supporting Information Table S3). Given that the local and total ln-transformed areas were highly correlated (Pearson's $r > .8$), only the former was retained in the subsequent analyses.

In addition, we estimated the current and LGM isolation as the minimum distance from the centroid of each alpine region to the boundary of the nearest alpine area ≥ 1,000 km^2. We set a threshold of 1,000 km^2 to exclude smaller alpine patches that could still be part of the same alpine region, that is, islands of the same archipelago (Steinbauer et al., 2016). If an alpine region had a total area ≥ 1,000 km^2, isolation was set to zero. Current and LGM alpine areas and isolation were ln-transformed.

Given that past climatic changes could have affected current diversity patterns (Graham et al., 2014), we also calculated the velocity of climate change since the LGM as a measure of regional climatic instability (Loarie et al., 2009; Sandel et al., 2011), using the "gVoCC" function of the VoCC package (Molinos et al., 2019) with current and LGM mean annual temperatures. The latter was calculated as the average of the two PMIP3 climatic datasets derived using the CCSM4 and MIROC-ESM climate models (Sandel et al., 2011; Weigelt et al., 2013, 2016).

Finally, we included biogeographical realms (Keil & Chase, 2019) as a proxy for differences in evolutionary history. Owing to the lack of regional data, we did not account for differences in the geological history of mountains. However, we acknowledge that this could influence speciation and partly explain species richness (Whittaker et al., 2008, 2017).

2.5 | Statistical analyses

We tested the influence of environment, geography and history on estimated regional richness by fitting individual Poisson generalized linear mixed-effects models (GLMMs) to each predictor with the "glmer" function of the R package lme4 (Bates et al., 2015). Initially, we tested univariate relationships to select a set of significant variables to be used in subsequent multivariate modelling. To account for uncertainties in regional richness estimates, we weighted the observations by the inverse of their 95% confidence interval width. We also scaled the predictors by subtracting their mean and dividing by their standard deviation across the regions to ensure model convergence. To control for overdispersion and reduce the risk of type I errors, we added an observation-level random effect (OLRE) to the models, that is, a unique level of a random effect for each data point that models the extra-Poisson variation present in the data (Harrison, 2014). The ratios between the sum of squared Pearson residuals and the residual degrees of freedom of the fitted models with OLRE indicated no additional overdispersion. Next, we analysed the correlations among significant predictors with the Pearson correlation coefficient. We found that some of our regional variables were strongly correlated with one another (Supporting Information Figure S5), limiting our ability to distinguish partial contributions. However, we built alternative multivariate models by retaining only the significant, uncorrelated predictors. Finally, we checked for the presence of spatial autocorrelation in model residuals with the Moran's *I* test implemented in the "testSpatialAutocorrelation" function of the DHARMa package (Hartig, 2020) and found none (Supporting Information Table S4). We also fitted a null (intercept-only) model to compare the goodness of the fits. Models were compared using the Akaike Information Criterion corrected for small sample sizes (AICc) in addition to marginal and conditional R^2 (mR^2 and cR^2), calculated with the "r.squaredGLMM" function of the MuMIn package (Barton, 2019). Given that the only random effect in the models was an OLRE, $mR^2 = cR^2$.

We modelled community richness by fitting Poisson GLMMs including the environmental, geographical and historical predictors. Growing degree days and precipitation of the growing season were highly correlated with temperature and evapotranspiration of the growing season, respectively (Pearson's $r > .6$). Likewise, area and isolation-related variables were highly correlated with one another. Thus, to avoid multicollinearity issues, we retained temperature and evapotranspiration of the growing season, in addition to current and LGM local areas, and excluded the other variables. We also accounted for different plot sizes by adding their ln-transformed values to the model and controlled for overdispersion by adding an OLRE. Given that the plot-level predictors were derived from digital datasets at 1 km resolution, we randomly selected one plot for each .01° cell (*c*. 1 km). We repeated the process 999 times and obtained as many random subsets of 2,534 plots, that is, one plot for each .01° cell. Before modelling, all predictor variables were scaled to ensure model convergence. We then fitted the GLMMs to each of the 999 subsets. Given that the Moran's *I* tests highlighted strong spatial autocorrelation of the models' residuals, we re-fitted the models including random intercepts for .05 (≈5 km) and .1 (≈10 km) degree cells, which largely resolved the issue (Moran's $I \approx 0$; $p > .05$ for 75% of model fits). We also tested for regional effects by fitting another model to the 999 subsets, with an additional random intercept for regions. Finally, we averaged the fixed effect coefficients of the resulting models using weights based on their AICc with the "model.avg" function of the MuMIn package. The two resulting averaged models (with and without the random intercept for region) were compared using mean AICc, mR^2 and cR^2, obtained by calculating the weighted average of the respective indices for the 999 fits.

3 | RESULTS

According to sample-based rarefaction and extrapolation of regional richness (estimated for 180 plots), the richest alpine regions in this study were the Colombian and Ecuadorian Andes (Neotropics; 543 species), followed by the Pamir Mountains (Eastern Palaearctic; 497 species) and the Altai Mountains (Eastern Palaearctic; 486 species). At intermediate species richness levels, we found the Central and Eastern Alps (Western Palaearctic; 387 species), Sayan Mountains (Eastern

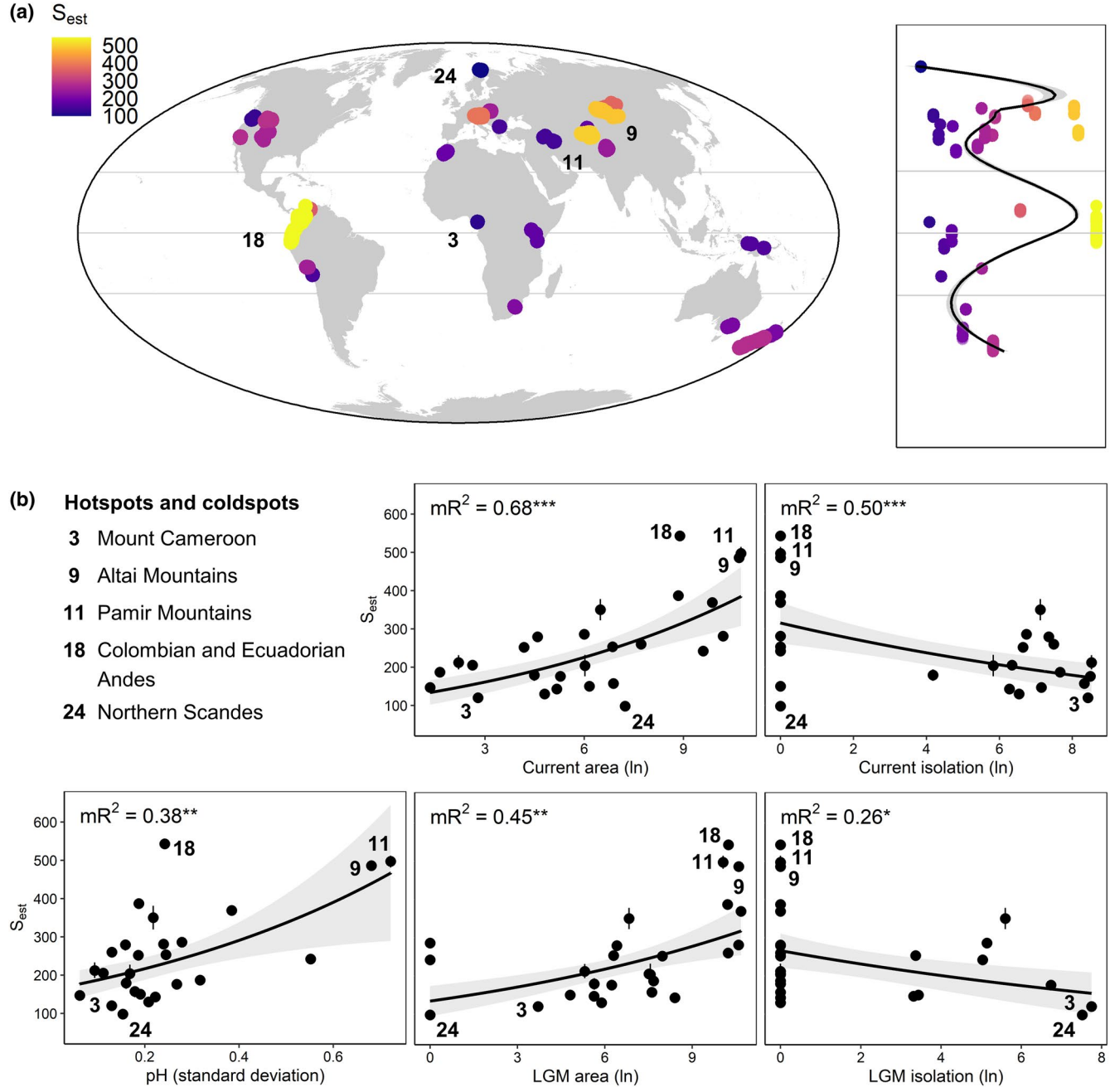


FIGURE 2 Latitudinal patterns and drivers of estimated regional species richness. (a) Regional plant species richness estimated for 180 plots (S_{est}). The scatterplot on the right represents the latitudinal trend. The three horizontal grey lines on the map and the scatterplot represent the equator and the tropics. The black line represents a generalized additive model (GAM) fit. (b) Single-predictor models of regional species richness. The dots represent the regional plant species richness estimated for 180 plots (S_{est}). The error bars represent the 95% bootstrap confidence intervals of the richness estimates. Black lines represent the individual Poisson generalized linear mixed-effects model (GLMM) fits. The grey bands are the 95% bootstrap confidence intervals. Marginal R^2 (mR^2) and model significance are reported. Significance codes: <.001 (***); <.01 (**); <.05 (*). The numbers of the main coldspot and hotspot regions are reported according to Figure 1. Abbreviation: LGM = Last Glacial Maximum [Colour figure can be viewed at wileyonlinelibrary.com]

Palaearctic; 369 species) and Cordillera de Mérida (Neotropics; 350 species). On the other end of the gradient, the poorest regions were Mount Cameroon (Afrotropics; 120 species) and the Northern Scandes (Western Palaearctic; 98 species) (Figure 1b,c).

The latitudinal pattern of regional richness peaked around the equator and at mid-latitudes of the Northern Hemisphere, with low species richness values around the tropics and at high latitudes (Figure 2a). At the community level, we observed a single peak at mid-latitudes of the Northern Hemisphere, but a wide range of species richness values across all regions (Figure 3a). The same patterns emerged even when different regional extents and plot sizes were accounted for, with an additional peak of regional richness at

mid-latitudes of the Southern Hemisphere corresponding to the Drakensberg (Afrotropics) and the Australian Alps (Australasia) (Supporting Information Figure S6a,b).

The null model of estimated regional richness had AICc = 67. The GLMMs of estimated species richness against individual predictors highlighted a positive significant effect of current area ($\hat{\beta}$ = .33 ± .07; AICc = 54), LGM area ($\hat{\beta}$ = .26 ± .08; AICc = 61) and soil pH variability ($\hat{\beta}$ = .24 ± .08; AICc = 61), whereas current isolation ($\hat{\beta}$ = −.26 ± .08; AICc = 61) and LGM isolation ($\hat{\beta}$ = −.20 ± .09; AICc = 65) had a negative effect (Figure 2b; Supporting Information Table S4; Figure S7). Among significant predictors, current area was correlated with soil pH variability (r = .70, p < .001) and with current isolation (r = −.76, p < .001), and LGM area was correlated with LGM isolation (r = −.72, p < .001) (Supporting Information Figure S5). Multivariate Poisson GLMMs with uncorrelated significant predictors explained 67% (current isolation + LGM isolation; AICc = 58) and 79% (current area + LGM area; AICc = 52) of the variance (Supporting Information Table S4).

At the community scale, species richness was positively related to the evapotranspiration of the growing season ($\hat{\beta}$ = .13 ± .05; p < .001), velocity of climate change ($\hat{\beta}$ = .10 ± .05; p < .001) and LGM alpine area ($\hat{\beta}$ = .19 ± .05; p < .001), whereas it was negatively related to soil pH ($\hat{\beta}$ = −.12 ± .05; p < .001). Nearctic plots were generally poorer in species than plots in other realms ($\hat{\beta}$ = −.44 ± .22; p < .001; Figure 3b; Supporting Information Table S5). Overall, the fixed effects explained 22% of the variance, and the inclusion of the random effects controlling for the spatial aggregation of plots at 5 and 10 km increased the explained variance to 58%. The inclusion of regions as an additional random effect increased the total explained variance further to 65% and left as significant fixed effects the mean temperature of the growing season ($\hat{\beta}$ = .04 ± .03; p < .05) and soil

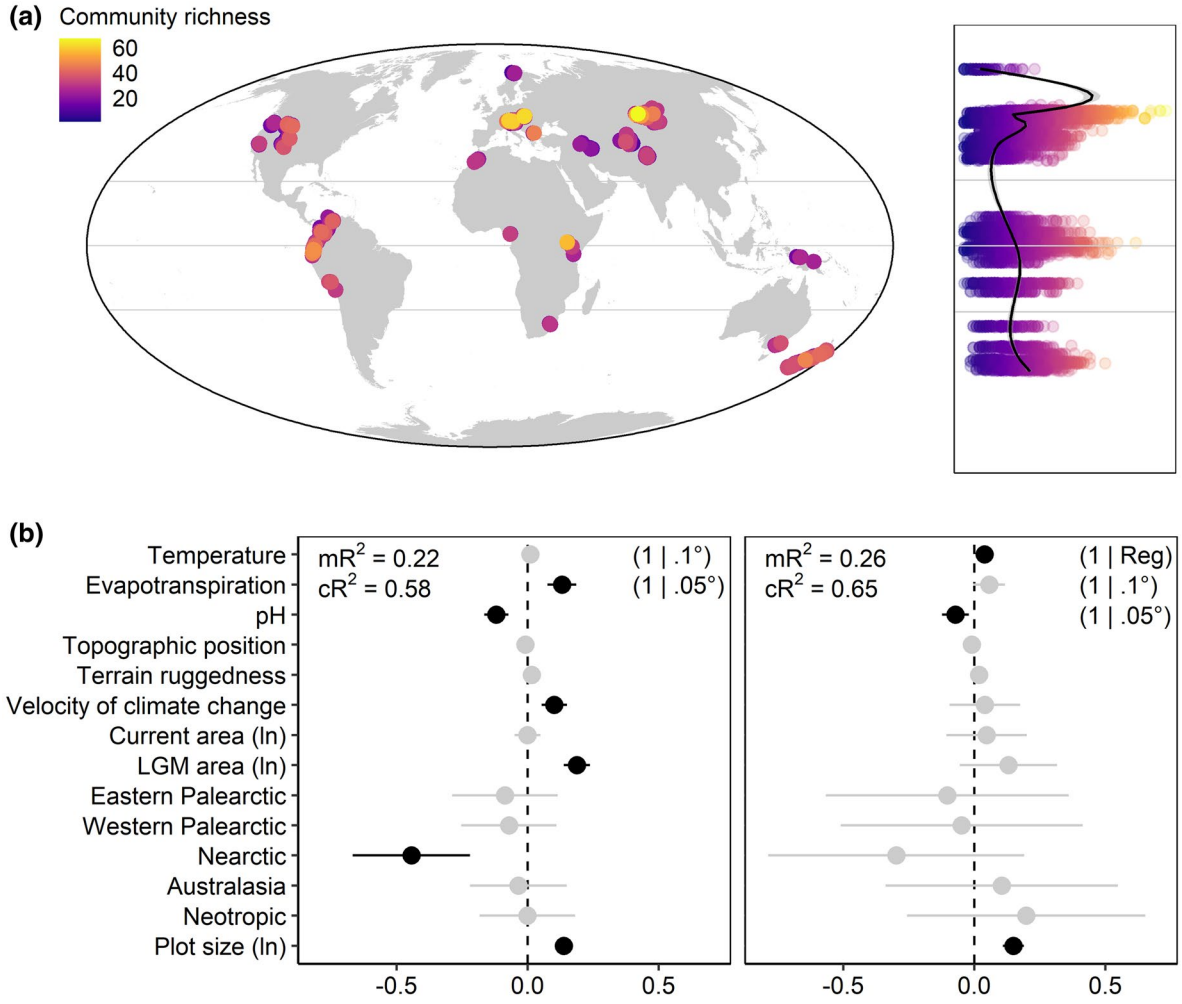


FIGURE 3 Latitudinal patterns and drivers of community species richness. (a) Community plant species richness. The scatterplot on the right represents the latitudinal trend. The three horizontal grey lines on the map and the scatterplot represent the equator and the tropics. The black line represents a generalized additive model (GAM) fit. (b) Standardized fixed-effect coefficients from model-averaged Poisson generalized linear mixed-effects models (GLMMs) of 999 sub-models of community species richness, based on 999 random subsets of 2,534 plots. Temperature and evapotranspiration are the mean values calculated during the growing season. Random effects: (1 | .05°) and (1 | .1°) group plots belonging to the same .05° and .1° cells; (1 | Reg) groups plots belonging to the same region. Dots and bars represent the mean and the 95% confidence interval of the coefficients. Significant coefficients are drawn in black. Marginal (mR^2) and conditional R^2 (cR^2) represent the weighted average of the corresponding measures across all sub-models. Abbreviation: LGM = Last Glacial Maximum [Colour figure can be viewed at wileyonlinelibrary.com]

pH ($\hat{\beta} = -.07 \pm .05$; $p < .05$), which together explained 26% of the variance (Figure 3b; Supporting Information Table S5).

4 | DISCUSSION

4.1 | Regional patterns and drivers

Our results, based on a representative sample of global alpine vegetation, showed a latitudinal pattern of plant species richness, with peaks at mid-latitudes and around the equator. The highest estimate of regional richness was detected in the Colombian and Ecuadorian Andes (Neotropics). This region is home to the páramo ecosystem, a centre of plant diversity within the tropical biodiversity hotspot known to host the richest alpine flora in the world (Madriñán et al., 2013; Myers et al., 2000). At higher latitudes, we also found that the Pamir and Altai Mountains (Eastern Palaearctic) exhibited regional richness comparable to the páramos, representing hotspots of alpine plant diversity outside the tropics. This is consistent with previous studies that highlighted the high plant diversity of the Altai, Pamir, in addition to other Central Asian mountain systems (Agakhanjanz & Breckle, 1995; Brummitt et al., 2020; Kier et al., 2005; Körner, 1995; Xing & Ree, 2017). Nevertheless, these putative mid-latitude alpine hotspots are generally excluded from global centres of biodiversity, despite their importance as refugia for cold-adapted plant species (Chytrý et al., 2019). When accounting for the extent of the local alpine area, the Drakensberg (Afrotropics) and the Australian Alps (Australasia) emerged as centres of alpine plant richness of the Southern Hemisphere. Indeed, the high-elevation plateau of the Drakensberg has been widely recognized as a continental hotspot of botanical diversity (Brand et al., 2019; Carbutt, 2019), and the Australian Alps have been listed among the main national areas of plant species richness (Bell et al., 2018; Crisp et al., 2001). Other regions showed lower species richness, with no clear distinctions among different realms or latitudes. The lowest richness was found in Mount Cameroon (Afrotropic) and the Northern Scandes (Western Palaearctic) which, interestingly, are located at the extremes of the latitudinal distribution of the global alpine biomes, one near the equator and the other at the Arctic Circle. Mount Cameroon represents the only isolated outpost of the alpine life zone in tropical Western Africa (Anthelme & Dangles, 2012) and is an active volcano, with eruptions that limit development of vegetation (Nagy & Grabherr, 2009). Moreover, the Northern Scandes were completely glaciated during the Pleistocene glacial maxima and located far from the Southern European glacial refugia further south (Lenoir et al., 2010).

Our models showed that the regional richness of alpine ecosystems is mostly independent of macroclimatic gradients. An analogous decoupling pattern has also been reported for the global diversity of grasses as a response to biogeographical history and the adaptation of certain lineages to cold and arid environments (Visser et al., 2014). Indeed, global alpine areas are climatically constrained toward low-temperature conditions (Körner, 2003; Nagy & Grabherr, 2009; Paulsen & Körner, 2014; Testolin et al., 2020). Thus, although alpine plants respond to changes in temperature and light because of topography, large-scale richness patterns of alpine vegetation seem to be largely independent of energy gradients that determine species diversity at lower elevations (Hillebrand, 2004). Moreover, global alpine areas are subjected to different amounts of precipitation and are differentiated along a gradient of humidity (Körner, 2003; Nagy & Grabherr, 2009; Testolin et al., 2020). Although our dataset encompasses a large portion of the variation in water availability of global alpine areas (Supporting Information Figure S1), the effect of precipitation on regional richness was not significant. This suggests that the association of water availability with plant species richness might be restricted to local scales and especially to arid regions, where precipitation is the main factor limiting plant growth (Palpurina et al., 2017).

Contrarily to macroclimate, we found a positive effect of the extent of current alpine area and a negative effect of isolation. The importance of these factors is consistent with the predictions of the theory of island biogeography (MacArthur & Wilson, 1967), which posits that larger, less isolated islands are characterized by lower extinction rates and greater chances of being colonized by new species. Nevertheless, the historical legacy of the extent and isolation of alpine areas during the LGM also left a strong imprint on regional richness patterns that is independent of their current geographical characteristics. The extent of alpine areas during the LGM was the second strongest predictor of regional richness and, together with the current area, explained almost 80% of the variance. This is consistent with recent refinements of the theory of island biogeography that incorporate the effect of Late Quaternary climate oscillations on oceanic islands (Fernández-Palacios et al., 2016; Weigelt et al., 2016). Pleistocene glacial–interglacial cycles acted like a "historical sieve" (Körner, 1995) on alpine plant diversity. During glacial periods, downslope shifts of the alpine belt resulted in increased surface area and connectivity of tropical alpine archipelagos, in addition to colonization and diversification processes in mid-latitude mountain ranges that favoured *in situ* speciation (Flantua et al., 2020). The high species richness found in the Andes is probably the result of multiple contingencies related to South American tropical diversity and strong past connectivity of these mountains (Flantua et al., 2019), which are not co-occurring in any other tropical region. Moreover, in Central Asian mountains, the emergence of habitat corridors during glacial periods resulted in extensive long-distance dispersal, with the consequent admixture of previously isolated floras (Agakhanjanz & Breckle, 1995; Agakhanyantz & Lopatin, 1978). Indeed, the Pamir Mountains are a continental hub for floristic migrations (the Pamir Knot) that connects south-central Asian ranges to the northern Siberian mountains (Agakhanjanz & Breckle, 1995), whereas the Altai Mountains connect diversity between Euro-Siberian and Central Asian floristic regions (Chytrý et al., 2012). Our results also show a positive relationship between regional richness and soil pH variability (a surrogate for bedrock heterogeneity) largely driven by the Pamir and Altai Mountains. This finding confirms the effect of habitat heterogeneity on species richness inherent to larger areas (Lomolino, 2000a) through the occurrence of more diverse bedrock types (Moser et al., 2005).

4.2 | Community patterns and drivers

At the community scale, the latitudinal pattern of species richness was less pronounced than at the regional scale, with a single peak at mid-latitudes of the Northern Hemisphere, but a wide range of values within all regions. While controlling for plot size, we found a positive effect of evapotranspiration of the growing season and a negative effect of soil pH on community richness. The former is consistent with the species–energy hypothesis, which states that more productive communities (i.e., where higher temperatures and solar radiation support greater photosynthetic rates) are also richer in species (Wright, 1983). The latter could be explained by the absence of strongly acidic soils (pH < 4) in our dataset. Furthermore, soils with high pH values can be linked to reduced nutrient availability in harsh conditions and the confounding effect of reduced precipitation (Chytrý et al., 2007; Palpurina et al., 2017), explaining the lower species richness in our dataset. Despite the underlying causes of these effects, our results are in line with the role of energy-driven processes and bedrock mineralogy as determinants of vascular plant species richness in alpine communities (Lenoir et al., 2010; Moser et al., 2005; Vonlanthen et al., 2006). Nevertheless, evolutionary and historical factors might also affect current patterns of community richness (Ricklefs & He, 2016). The Nearctic realm exhibited lower community richness than any other biogeographical realm, possibly owing to limited evolutionary radiation of the North American temperate flora (Qian & Ricklefs, 2000). In addition, the velocity of climate change and the LGM extent of alpine area were both positively related to community richness, indicating that the greater availability of alpine habitats in the past influences plant species richness at the community scale (Pärtel & Zobel, 1999).

Large-scale environmental factors, however, explained only a limited proportion of the variation in community richness compared with regional and sub-regional effects, suggesting that dispersal-related processes and other spatially structured factors strongly influence local richness patterns (Dormann et al., 2007). In alpine landscapes, these effects are regulated by elevational and meso-topographical gradients that affect microclimatic conditions and local plant diversity (Bruun et al., 2006; Jiménez-Alfaro et al., 2014; Scherrer & Körner, 2011). A weak influence of global macroclimatic gradients on local communities has also been detected for functional diversity across plant formations (Bruelheide et al., 2018), but it had not been tested before on a single ecosystem. Our results, therefore, support the dominant role of within-region effects linked to postglacial spatial configuration and historical contingencies, rather than macroclimatic factors, when explaining the global variation of alpine local communities.

4.3 | Data constraints and assumptions

Despite including several alpine regions across all continents and latitudes, our dataset lacked information about some outstanding centres of alpine plant diversity, such as the Himalayas and Hengduan Mountains (Ding et al., 2020; Favre et al., 2015; Muellner-Riehl et al., 2019; Xing & Ree, 2017) or the Caucasus (Agakhanjanz & Breckle, 1995; Körner, 1995), owing to the limited availability of community data from these areas. Therefore, our results related to the latitudinal patterns of regional and community richness could be refined further by the future addition of data from currently missing regions. Nevertheless, our aim was not to present a complete census of global alpine regions, but to assess their richness patterns and the corresponding drivers using a representative sample. In this respect, the collection of georeferenced vegetation plots presented here encompasses all continents and a wide range of latitudes; it represents regions with markedly different biogeographical history and vegetation types growing on different substrates, and covers a large portion of the climatic envelope of global alpine areas (Supporting Information Figure S1). Despite the lack of some remarkable alpine regions, our dataset allowed us to highlight the presence of extratropical alpine diversity centres and the importance of historical factors in shaping the current alpine plant richness patterns.

We also note that the use of heterogeneous surveys from different collectors might create issues related to different sampling effort among regions. We controlled for sampling effort in two ways. First, we used rarefaction and extrapolation techniques that assumed that the spatial distribution of plots in each region was representative of the regional diversity. Although this assumption is difficult to prove without additional data, we note that our regional samples were selected to capture the local heterogeneity of vegetation types and covered a wide range of elevations in all regions (Supporting Information Table S3), thus increasing the probability that our regional richness estimates were correlated with regional species pools. Second, we explicitly quantified the proportion of the alpine area sampled in each region, thus allowing inter-regional comparisons even when the samples covered very different extents or only a small fraction of the total available alpine area (e.g., Ladakh Range, Pamir Mountains or Southern Cordillera Occidental Peru).

In addition, we note that taxon concepts might not be applied consistently across all datasets, that is, they are the result of "lumping" and "splitting" of taxa delimitations that change with time and from place to place (Rouhan & Gaudeul, 2014; Wiser, 2016). Although a harmonized species nomenclature cannot account fully for this taxonomic bias (Wiser, 2016), it still represents the most effective tool to address taxonomic inflation in macroecological studies (Isaac et al., 2004). Indeed, by correcting misspelt names and merging synonyms, we assume that the main sources of error relevant to the estimation of species richness in different regions were removed, and remaining issues about potential pitfalls in the geographical distribution of species (Boyle et al., 2013) are not pertinent to the present study.

4.4 | Conclusions

Overall, we found that the latitudinal distribution of plant species richness in alpine ecosystems is decoupled from the general latitudinal diversity gradient and that it relates to regional idiosyncrasies,

rather than macroclimatic gradients. Although our results are conclusive enough to support that current and historical effects of area, isolation and environmental heterogeneity exert an overarching influence on vascular plant richness in global alpine ecosystems, we are still far from understanding the processes behind such effects. Future alpine research should, therefore, consider local information about soil biotic and abiotic composition, topographical features and microclimatic variation at the regional scale. Additionally, further efforts should be oriented toward the collection of plant community data from underrepresented regions. Indeed, this work is the starting point for defining global hotspots of alpine plant diversity, and further investigations including patterns of endemism, functional variation and phylogenetic diversity are still needed. This type of information, together with dynamic regional diversity models accounting for spatio-temporal connectivity, will provide a better understanding of the patterns we have found here and a tool for the effective conservation of alpine biodiversity in response to climate change.

ACKNOWLEDGMENTS

We thank Keith McDougall (Department of Planning, Industry and Environment, Queanbeyan, NSW, Australia) for providing the vegetation data for the Australian Alps and Miska Luoto (Department of Geosciences and Geography, University of Helsinki, Helsinki, Finland) for sampling part of the vegetation data in the Scandinavian Mountains, along with all the other data collectors. B.J.-A. was funded by the Marie Curie Clarín-COFUND program of the Principality of Asturias-EU (ACB17-26) and the Spanish Research Agency (AEI/10.13039/501100011033). M.C. was funded by the Czech Science Foundation (project no. 19-28491X). S.P.S. was funded by the Leverhulme Trust (UK), the University of Zurich, the Swiss National Science Foundation (grant P2ZHP3_161988) and the Deutsche Forschungsgemeinschaft (grants HE 6726 5/1; JU 2748/6-1). G.P.M. was funded by U.S. National Science Foundation award 1853665. J.D. was funded by the Ministerstvo Školství, Mládeže a Tělovýchovy Inter-Excellence project (LTAUSA18007). S.K.W. was funded by the New Zealand Ministry for Business, Innovation and Employment's Strategic Science Investment Fund. We also appreciate the support of the Deutsche Forschungsgemeinschaft for funding sPlot - The Global Vegetation Database as one of the German Centre for Integrative Biodiversity Research (iDiv) (DFG FZT 118, 202548816) platforms, and the organization of three workshops through the Synthesis Centre for Biodiversity Sciences (sDiv) calls. H.B., S.H., U.J. and F.M.S. appreciate support by iDiv. This work is part of ALPVEG, a research network on alpine vegetation ecology (www.alpveg.com).

AUTHOR CONTRIBUTIONS

R.T. and B.J.-A. conceived the study and developed the methodology. P.B., R.F.B., J.D., M.F., A.H., M.K., A.Y.K., J.L., N.M., G.P.M., D.M.-T., J.N., A.N., R.K.P., G.P., J.Š., P.S., S.P.S., K.V., R.V., W.W., S.K.W. and E.G.Z. provided the vegetation plot data. F.M.S. and H.B. facilitated access to the sPlot database. R.T. analysed the data and produced the outputs. R.T. and B.J.-A. wrote the first manuscript draft. B.J.-A. supervised the study. All the authors discussed the methodology and commented on various versions of the manuscript.

DATA AVAILABILITY STATEMENT

The data and R code used to produce the main results presented in the article, in addition to the link to an online interactive map displaying the location of the plots, are available at: https://dx.doi.org/10.6084/m9.figshare.14191268

ORCID

Riccardo Testolin https://orcid.org/0000-0002-8916-7231
Fabio Attorre https://orcid.org/0000-0002-7744-2195
Robert F. Brand https://orcid.org/0000-0002-9790-3120
Helge Bruelheide https://orcid.org/0000-0003-3135-0356
Milan Chytrý https://orcid.org/0000-0002-8122-3075
Michele De Sanctis https://orcid.org/0000-0002-7280-6199
Jiri Dolezal https://orcid.org/0000-0002-5829-4051
Manfred Finckh https://orcid.org/0000-0003-2186-0854
Sylvia Haider https://orcid.org/0000-0002-2966-0534
Andreas Hemp https://orcid.org/0000-0002-5369-2122
Ute Jandt https://orcid.org/0000-0002-3177-3669
Michael Kessler https://orcid.org/0000-0003-4612-9937
Jonathan Lenoir https://orcid.org/0000-0003-0638-9582
George P. Malanson https://orcid.org/0000-0001-9527-0086
Daniel B. Montesinos-Tubée https://orcid.org/0000-0002-4439-5089
Jalil Noroozi https://orcid.org/0000-0003-4124-2359
Arkadiusz Nowak https://orcid.org/0000-0001-8638-0208
Robert K. Peet https://orcid.org/0000-0003-2823-6587
Gwendolyn Peyre https://orcid.org/0000-0002-1977-7181
Francesco Maria Sabatini https://orcid.org/0000-0002-7202-7697
Jozef Šibík https://orcid.org/0000-0002-5949-862X
Petr Sklenář https://orcid.org/0000-0003-0429-2621
Steven P. Sylvester https://orcid.org/0000-0001-5577-8782
Kiril Vassilev https://orcid.org/0000-0003-4376-5575
Risto Virtanen https://orcid.org/0000-0002-8295-8217
Wolfgang Willner https://orcid.org/0000-0003-1591-8386
Susan K. Wiser https://orcid.org/0000-0002-8938-8181
Evgeny G. Zibzeev https://orcid.org/0000-0002-7135-0724
Borja Jiménez-Alfaro https://orcid.org/0000-0001-6601-9597

REFERENCES

Agakhanjanz, O., & Breckle, S.-W. (1995). Origin and evolution of the mountain flora in middle Asia and neighbouring mountain regions. In F. S. Chapin III, & C. Körner (Eds.), *Arctic and alpine biodiversity: Patterns, causes and ecosystem consequences. Ecological Studies (Analysis and Synthesis)* (Vol. *113*, pp. 63–80). Springer. https://doi.org/10.1007/978-3-642-78966-3_5

Agakhanyantz, O. E., & Lopatin, I. K. (1978). Main characteristics of the ecosystems of the Pamirs, USSR. *Arctic and Alpine Research*, *10*, 397. https://doi.org/10.2307/1550770

Amatulli, G., Domisch, S., Tuanmu, M. N., Parmentier, B., Ranipeta, A., Malczyk, J., & Jetz, W. (2018). A suite of global, cross-scale topographic variables for environmental and biodiversity modeling. *Scientific Data*, *5*, 1080040. https://doi.org/10.1038/sdata.2018.40

Anthelme, F., & Dangles, O. (2012). Plant–plant interactions in tropical alpine environments. *Perspectives in Plant Ecology, Evolution and Systematics*, *14*, 363–372. https://doi.org/10.1016/j.ppees.2012.05.002

Antonelli, A., Kissling, W. D., Flantua, S. G. A., Bermúdez, M. A., Mulch, A., Muellner-Riehl, A. N., Kreft, H., Linder, H. P., Badgley, C., Fjeldså, J., Fritz,

S. A., Rahbek, C., Herman, F., Hooghiemstra, H., & Hoorn, C. (2018). Geological and climatic influences on mountain biodiversity. *Nature Geoscience*, *11*, 718–725. https://doi.org/10.1038/s41561-018-0236-z

Barton, K. (2019). *MuMIn: Multi-model inference*. R package version 1.43.15. https://cran.r-project.org/package=MuMIn

Bates, D., Mächler, M., Bolker, B. M., & Walker, S. C. (2015). Fitting linear mixed-effects models using lme4. *Journal of Statistical Software*, *67*, 1–48. https://doi.org/10.18637/jss.v067.i01

Beguería, S., & Vicente-Serrano, S. M. (2017). *SPEI: Calculation of the standardised precipitation-evapotranspiration index*. R package version 1.7. https://cran.r-project.org/package=SPEI%0A

Bell, N., Griffin, P. C., Hoffmann, A. A., & Miller, A. D. (2018). Spatial patterns of genetic diversity among Australian alpine flora communities revealed by comparative phylogenomics. *Journal of Biogeography*, *45*, 177–189. https://doi.org/10.1111/jbi.13120

Boyle, B., Hopkins, N., Lu, Z., Raygoza Garay, J. A., Mozzherin, D., Rees, T., Matasci, N., Narro, M. L., Piel, W. H., Mckay, S. J., Lowry, S., Freeland, C., Peet, R. K., & Enquist, B. J. (2013). The taxonomic name resolution service: An online tool for automated standardization of plant names. *BMC Bioinformatics*, *14*, 16. https://doi.org/10.1186/1471-2105-14-16

Brand, R. F., Collins, N., & Du Preez, P. J. (2015). A phytosociology survey and vegetation description of inselbergs in the uKhahlamba-Drakensberg Park World Heritage Site, South Africa. *Koedoe*, *57*, a1233. https://doi.org/10.4102/koedoe.v57i1.1233

Brand, R. F., Scott-Shaw, C. R., & O'Connor, T. G. (2019). The alpine flora on inselberg summits in the Maloti-Drakensberg Park, KwaZulu-Natal, South Africa. *Bothalia*, *49*, 1–15. https://doi.org/10.4102/abc.v49i1.2386

Bruelheide, H., Dengler, J., Jiménez-Alfaro, B., Purschke, O., Hennekens, S. M., Chytrý, M., Pillar, V. D., Jansen, F., Kattge, J., Sandel, B., Aubin, I., Biurrun, I., Field, R., Haider, S., Jandt, U., Lenoir, J., Peet, R. K., Peyre, G., Sabatini, F. M., … Zverev, A. (2019). sPlot—A new tool for global vegetation analyses. *Journal of Vegetation Science*, *30*, 161–186. https://doi.org/10.1111/jvs.12710

Bruelheide, H., Dengler, J., Purschke, O., Lenoir, J., Jiménez-Alfaro, B., Hennekens, S. M., Botta-Dukát, Z., Chytrý, M., Field, R., Jansen, F., Kattge, J., Pillar, V. D., Schrodt, F., Mahecha, M. D., Peet, R. K., Sandel, B., van Bodegom, P., Altman, J., Alvarez-Dávila, E., … Jandt, U. (2018). Global trait–environment relationships of plant communities. *Nature Ecology and Evolution*, *2*, 1906–1917. https://doi.org/10.1038/s41559-018-0699-8

Brummitt, N., Araújo, A. C., & Harris, I. (2020). Areas of plant diversity — What do we know? *Plants, People, Planet*, *3*, 33–44. https://doi.org/10.1002/ppp3.10110

Bruun, H. H., Moen, J., Virtanen, R., Grytnes, J.-A., Oksanen, L., & Angerbjörn, A. (2006). Effects of altitude and topography on species richness of vascular plants, bryophytes and lichens in alpine communities. *Journal of Vegetation Science*, *17*, 37–46. https://doi.org/10.1111/j.1654-1103.2006.tb02421.x

Carbutt, C. (2019). The Drakensberg Mountain Centre: A necessary revision of southern Africa's high-elevation centre of plant endemism. *South African Journal of Botany*, *124*, 508–529. https://doi.org/10.1016/j.sajb.2019.05.032

Chao, A., Gotelli, N. J., Hsieh, T. C., Sander, E. L., Ma, K. H., Colwell, R. K., & Ellison, A. M. (2014). Rarefaction and extrapolation with Hill numbers: A framework for sampling and estimation in species diversity studies. *Ecological Monographs*, *84*, 45–67. https://doi.org/10.1890/13-0133.1

Chytrý, M., Danihelka, J., Ermakov, N., Hájek, M., Hájková, P., Kočí, M., Kubešová, S., Lustyk, P., Otýpková, Z., Popov, D., Roleček, J., Řezníčková, M., Šmarda, P., & Valachovič, M. (2007). Plant species richness in continental southern Siberia: Effects of pH and climate in the context of the species pool hypothesis. *Global Ecology and Biogeography*, *16*, 668–678. https://doi.org/10.1111/j.1466-8238.2007.00320.x

Chytrý, M., Ermakov, N., Danihelka, J., Hájek, M., Hájková, P., Horsák, M., Kočí, M., Kubešová, S., Lustyk, P., Otýpková, Z., Pelánková, B., Valachovič, M., & Zelený, D. (2012). High species richness in hemiboreal forests of the northern Russian Altai, southern Siberia. *Journal of Vegetation Science*, *23*, 605–616. https://doi.org/10.1111/j.1654-1103.2011.01383.x

Chytrý, M., Horsák, M., Danihelka, J., Ermakov, N., German, D. A., Hájek, M., Hájková, P., Kočí, M., Kubešová, S., Lustyk, P., Nekola, J. C., Pavelková Řičánková, V., Preislerová, Z., Resl, P., & Valachovič, M. (2019). A modern analogue of the Pleistocene steppe-tundra ecosystem in southern Siberia. *Boreas*, *48*, 36–56. https://doi.org/10.1111/bor.12338

Crisp, M. D., Laffan, S., Linder, H. P., & Monro, A. (2001). Endemism in the Australian flora. *Journal of Biogeography*, *28*, 183–198. https://doi.org/10.1046/j.1365-2699.2001.00524.x

Ding, W. N., Ree, R. H., Spicer, R. A., & Xing, Y. W. (2020). Ancient orogenic and monsoon-driven assembly of the world's richest temperate alpine flora. *Science*, *369*, 578–581. https://doi.org/10.1126/science.abb4484

Dormann, C. F., McPherson, J. M., Araújo, M. B., Bivand, R., Bolliger, J., Carl, G., Davies, R. G., Hirzel, A., Jetz, W., Daniel Kissling, W., Kühn, I., Ohlemüller, R., Peres-Neto, P. R., Reineking, B., Schröder, B., Schurr, F. M., & Wilson, R. (2007). Methods to account for spatial autocorrelation in the analysis of species distributional data: A review. *Ecography*, *30*, 609–628. https://doi.org/10.1111/j.2007.0906-7590.05171.x

Farr, T. G., Rosen, P. A., Caro, E., Crippen, R., Duren, R., Hensley, S., Kobrick, M., Paller, M., Rodriguez, E., Roth, L., Seal, D., Shaffer, S., Shimada, J., Umland, J., Werner, M., Oskin, M., Burbank, D., & Alsdorf, D. (2007). The shuttle radar topography mission. *Reviews of Geophysics*, *45*, 1–33. https://doi.org/10.1029/2005RG000183

Favre, A., Päckert, M., Pauls, S. U., Jähnig, S. C., Uhl, D., Michalak, I., & Muellner-Riehl, A. N. (2015). The role of the uplift of the Qinghai-Tibetan Plateau for the evolution of Tibetan biotas. *Biological Reviews of the Cambridge Philosophical Society*, *90*, 236–253. https://doi.org/10.1111/brv.12107

Fernández-Palacios, J. M., Rijsdijk, K. F., Norder, S. J., Otto, R., Nascimento, L., Fernández-Lugo, S., Tjørve, E., Whittaker, R. J., & Santos, A. (2016). Towards a glacial-sensitive model of island biogeography. *Global Ecology and Biogeography*, *25*, 817–830. https://doi.org/10.1111/geb.12320

Flantua, S. G. A., O'Dea, A., Onstein, R. E., Giraldo, C., & Hooghiemstra, H. (2019). The flickering connectivity system of the north Andean páramos. *Journal of Biogeography*, *46*, 1808–1825. https://doi.org/10.1111/jbi.13607

Flantua, S. G. A., Payne, D., Borregaard, M. K., Beierkuhnlein, C., Steinbauer, M. J., Dullinger, S., Essl, F., Irl, S. D. H., Kienle, D., Kreft, H., Lenzner, B., Norder, S. J., Rijsdijk, K. F., Rumpf, S. B., Weigelt, P., & Field, R. (2020). Snapshot isolation and isolation history challenge the analogy between mountains and islands used to understand endemism. *Global Ecology and Biogeography*, *29*, 1651–1673. https://doi.org/10.1111/geb.13155

Gottfried, M., Pauli, H., Futschik, A., Akhalkatsi, M., Barančok, P., Benito Alonso, J. L., Coldea, G., Dick, J., Erschbamer, B., Fernández Calzado, M. R., Kazakis, G., Krajči, J., Larsson, P., Mallaun, M., Michelsen, O., Moiseev, D., Moiseev, P., Molau, U., Merzouki, A., … Grabherr, G. (2012). Continent-wide response of mountain vegetation to climate change. *Nature Climate Change*, *2*, 111–115. https://doi.org/10.1038/nclimate1329

Graham, C. H., Carnaval, A. C., Cadena, C. D., Zamudio, K. R., Roberts, T. E., Parra, J. L., McCain, C. M., Bowie, R. C. K., Moritz, C., Baines, S. B., Schneider, C. J., VanDerWal, J., Rahbek, C., Kozak, K. H., & Sanders, N. J. (2014). The origin and maintenance of montane diversity: Integrating evolutionary and ecological processes. *Ecography*, *37*, 711–719. https://doi.org/10.1111/ecog.00578

Harris, S. A. (2007). Biodiversity of the alpine vascular flora of the N.W. North American Cordillera: The evidence from phyto-geography. *Erdkunde*, *61*, 344–357. https://doi.org/10.3112/erdkunde.2007.04.05

Harrison, X. A. (2014). Using observation-level random effects to model overdispersion in count data in ecology and evolution. *PeerJ*, *2*, e616. https://doi.org/10.7717/peerj.616

Hartig, F. (2020). *DHARMa: Residual diagnostics for hierarchical (multi-level/mixed) regression models*. R package version 0.3.1. https://cran.r-project.org/package=DHARMa

Heaney, L. R. (2000). Dynamic disequilibrium: A long-term, large-scale perspective on the equilibrium model of island biogeography. *Global Ecology and Biogeography*, *9*, 59–74. https://doi.org/10.1046/j.1365-2699.2000.00163.x

Hengl, T., Mendes de Jesus, J., Heuvelink, G. B. M., Ruiperez Gonzalez, M., Kilibarda, M., Blagotić, A., Shangguan, W., Wright, M. N., Geng, X., Bauer-Marschallinger, B., Guevara, M. A., Vargas, R., MacMillan, R. A., Batjes, N. H., Leenaars, J. G. B., Ribeiro, E., Wheeler, I., Mantel, S., & Kempen, B. (2017). SoilGrids250m: Global gridded soil information based on machine learning. *PLoS One*, *12*, e0169748. https://doi.org/10.1371/journal.pone.0169748

Hillebrand, H. (2004). On the generality of the latitudinal diversity gradient. *The American Naturalist*, *163*, 192–211. https://doi.org/10.1086/381004

Hope, G. S., Peterson, J. A., Allison, I., & Uwe, R. (1976). *The equatorial glaciers of New Guinea*. A. A. Balkema.

Hsieh, T. C., Ma, K. H., & Chao, A. (2016). iNEXT: An R package for rarefaction and extrapolation of species diversity (Hill numbers). *Methods in Ecology and Evolution*, *7*, 1451–1456. https://doi.org/10.1111/2041-210X.12613

Isaac, N. J. B., Mallet, J., & Mace, G. M. (2004). Taxonomic inflation: Its influence on macroecology and conservation. *Trends in Ecology and Evolution*, *19*, 464–469. https://doi.org/10.1016/j.tree.2004.06.004

Jiménez-Alfaro, B., Girardello, M., Chytrý, M., Svenning, J.-C., Willner, W., Gégout, J.-C., Agrillo, E., Campos, J. A., Jandt, U., Kącki, Z., Šilc, U., Slezák, M., Tichý, L., Tsiripidis, I., Turtureanu, P. D., Ujházyová, M., & Wohlgemuth, T. (2018). History and environment shape species pools and community diversity in European beech forests. *Nature Ecology and Evolution*, *2*, 483–490. https://doi.org/10.1038/s41559-017-0462-6

Jiménez-Alfaro, B., Marcenó, C., Bueno, Á., Gavilán, R., & Obeso, J. R. (2014). Biogeographic deconstruction of alpine plant communities along altitudinal and topographic gradients. *Journal of Vegetation Science*, *25*, 160–171. https://doi.org/10.1111/jvs.12060

Karger, D. N., Conrad, O., Böhner, J., Kawohl, T., Kreft, H., Soria-Auza, R. W., Zimmermann, N. E., Linder, H. P., & Kessler, M. (2017). Climatologies at high resolution for the earth's land surface areas. *Scientific Data*, *4*, 170122. https://doi.org/10.1038/sdata.2017.122

Keil, P., & Chase, J. M. (2019). Global patterns and drivers of tree diversity integrated across a continuum of spatial grains. *Nature Ecology and Evolution*, *3*, 390–399. https://doi.org/10.1038/s41559-019-0799-0

Kier, G., Mutke, J., Dinerstein, E., Ricketts, T. H., Küper, W., Kreft, H., & Barthlott, W. (2005). Global patterns of plant diversity and floristic knowledge. *Journal of Biogeography*, *32*, 1107–1116. https://doi.org/10.1111/j.1365-2699.2005.01272.x

Körner, C. (1995). Alpine plant diversity: A global survey and functional interpretations. In F. S. Chapin III, & C. Körner (Eds.), *Arctic and alpine biodiversity: Patterns, causes and ecosystem consequences* (pp. 45–62). Springer. https://doi.org/10.1007/978-1-4613-8395-6

Körner, C. (2003). *Alpine plant life. Functional plant ecology of high mountain ecosystems* (2nd ed.). Springer. https://doi.org/10.1007/978-3-642-18970-8

Körner, C., Jetz, W., Paulsen, J., Payne, D., Rudmann-Maurer, K., & M. Spehn, E. (2017). A global inventory of mountains for bio-geographical applications. *Alpine Botany*, *127*, 1–15. https://doi.org/10.1007/s00035-016-0182-6

Körner, C., Paulsen, J., & Spehn, E. M. (2011). A definition of mountains and their bioclimatic belts for global comparisons of biodiversity data. *Alpine Botany*, *121*, 73–78. https://doi.org/10.1007/s00035-011-0094-4

Kreft, H., & Jetz, W. (2007). Global patterns and determinants of vascular plant diversity. *Proceedings of the National Academy of Sciences USA*, *104*, 5925–5930. https://doi.org/10.1073/pnas.0608361104

Kreft, H., Jetz, W., Mutke, J., Kier, G., & Barthlott, W. (2008). Global diversity of island floras from a macroecological perspective. *Ecology Letters*, *11*, 116–127. https://doi.org/10.1111/j.1461-0248.2007.01129.x

Lenoir, J., Gégout, J.-C., Guisan, A., Vittoz, P., Wohlgemuth, T., Zimmermann, N. E., Dullinger, S., Pauli, H., Willner, W., Grytnes, J.-A., Virtanen, R., & Svenning, J.-C. (2010). Cross-scale analysis of the region effect on vascular plant species diversity in southern and northern European mountain ranges. *PLoS One*, *5*, 6–8. https://doi.org/10.1371/journal.pone.0015734

Loarie, S. R., Duffy, P. B., Hamilton, H., Asner, G. P., Field, C. B., & Ackerly, D. D. (2009). The velocity of climate change. *Nature*, *462*, 1052–1055. https://doi.org/10.1038/nature08649

Lomolino, M. V. (2000a). A call for a new paradigm of island biogeography. *Global Change Biology*, *9*, 1–6. https://doi.org/10.1046/j.1365-2699.2000.00185.x

Lomolino, M. V. (2000b). Ecology's most general, yet protean pattern: The species-area relationship. *Journal of Biogeography*, *27*, 17–26. https://doi.org/10.1046/j.1365-2699.2000.00377.x

Lomolino, M. V., Riddle, B. R., & Whittaker, R. J. (2017). *Biogeography: Biological diversity across space and time* (5th ed). Sinauer Associates.

MacArthur, R. H., & Wilson, E. O. (1967). *The theory of island biogeography*. Princeton University Press.

Madriñán, S., Cortés, A. J., & Richardson, J. E. (2013). Páramo is the world's fastest evolving and coolest biodiversity hotspot. *Frontiers in Genetics*, *4*, 192. https://doi.org/10.3389/fgene.2013.00192

McCormack, J. E., Huang, H., & Knowles, L. L. (2009). Sky islands. In R. G. Gillespie, & D. Clague (Eds.), *Encyclopedia of islands* (pp. 839–843). University of Chicago Press.

McGlone, M. S., Duncan, R. P., & Heenan, P. B. (2001). Endemism, species selection and the origin and distribution of the vascular plant flora of New Zealand. *Journal of Biogeography*, *28*, 199–216. https://doi.org/10.1046/j.1365-2699.2001.00525.x

Molinos, J. G., Schoeman, D. S., Brown, C. J., & Burrows, M. T. (2019). *VoCC: The velocity of climate change and related climatic metrics*. R package version 1.0.0. https://github.com/JorGarMol/VoCC%0A

Moser, D., Dullinger, S., Englisch, T., Niklfeld, H., Plutzar, C., Sauberer, N., Zechmeister, H. G., & Grabherr, G. (2005). Environmental determinants of vascular plant species richness in the Austrian Alps. *Journal of Biogeography*, *32*, 1117–1127. https://doi.org/10.1111/j.1365-2699.2005.01265.x

Muellner-Riehl, A. N., Schnitzler, J., Kissling, W. D., Mosbrugger, V., Rijsdijk, K. F., Seijmonsbergen, A. C., Versteegh, H., & Favre, A. (2019). Origins of global mountain plant biodiversity: Testing the "mountain-geobiodiversity hypothesis". *Journal of Biogeography*, *46*, 2826–2838. https://doi.org/10.1111/jbi.13715

Myers, N., Mittermeier, R. A., Mittermeier, C. G., da Fonseca, G. A. B., & Kent, J. (2000). Biodiversity hotspots for conservation priorities. *Nature*, *403*, 853–858. https://doi.org/10.1038/35002501

Nagy, L., & Grabherr, G. (2009). *The biology of alpine habitats*. Oxford University Press.

NASA, & JPL. (2013). *NASA shuttle radar topography mission global 1 arc second number [SRTMGL1N_003]*. NASA EOSDIS Land Processes DAAC. (Accessed 10 January 2019.)

Olson, D. M., Dinerstein, E., Wikramanayake, E. D., Burgess, N. D., Powell, G. V. N., Underwood, E. C., & Kassem, K. R. (2001). Terrestrial ecoregions of the world: A new map of life on earth. *BioScience*, *51*, 933–938.

Palpurina, S., Wagner, V., von Wehrden, H., Hájek, M., Horsák, M., Brinkert, A., Hölzel, N., Wesche, K., Kamp, J., Hájková, P., Danihelka, J., Lustyk, P., Merunková, K., Preislerová, Z., Kočí, M., Kubešová, S., Cherosov, M., Ermakov, N., German, D., … Chytrý, M. (2017). The relationship between plant species richness and soil pH vanishes with increasing aridity across Eurasian dry grasslands. *Global Ecology and Biogeography*, *26*, 425–434. https://doi.org/10.1111/geb.12549

Pärtel, M., & Zobel, M. (1999). Small-scale plant species richness in calcareous grasslands determined by the species pool, community age and shoot density. *Ecography*, *22*, 153–159. https://doi.org/10.1111/j.1600-0587.1999.tb00463.x

Pauli, H., Gottfried, M., Dullinger, S., Abdaladze, O., Akhalkatsi, M., Alonso, J. L. B., Coldea, G., Dick, J., Erschbamer, B., Calzado, R. F., Ghosn, D., Holten, J. I., Kanka, R., Kazakis, G., Kollar, J., Larsson, P., Moiseev, P., Moiseev, D., Molau, U., … Grabherr, G. (2012). Recent plant diversity changes on Europe's mountain summits. *Science*, *336*, 353–355. https://doi.org/10.1126/science.1219033

Paulsen, J., & Körner, C. (2014). A climate-based model to predict potential treeline position around the globe. *Alpine Botany*, *124*, 1–12. https://doi.org/10.1007/s00035-014-0124-0

Qian, H., & Ricklefs, R. E. (2000). Large-scale processes and the Asian bias in species diversity of temperate plants. *Nature*, *407*, 180–182. https://doi.org/10.1038/35025052

Quinn, J. A. (2008). *Arctic and alpine biomes*. Greenwood.

R Core Team. (2020). *R: A language and environment for statistical computing*. R Foundation for Statistical Computing. https://www.r-project.org/

Rahbek, C., Borregaard, M. K., Antonelli, A., Colwell, R. K., Holt, B. G., Nogues-Bravo, D., Rasmussen, C. M. Ø., Richardson, K., Rosing, M. T., Whittaker, R. J., & Fjeldså, J. (2019). Building mountain biodiversity: Geological and evolutionary processes. *Science*, *365*, 1114–1119. https://doi.org/10.1126/science.aax0151

Rahbek, C., Borregaard, M. K., Colwell, R. K., Dalsgaard, B. O., Holt, B. G., Morueta-Holme, N., Nogues-Bravo, D., Whittaker, R. J., & Fjeldså, J. (2019). Humboldt's enigma: What causes global patterns of mountain biodiversity? *Science*, *365*, 1108–1113. https://doi.org/10.1126/science.aax0149

Ricklefs, R. E., & He, F. (2016). Region effects influence local tree species diversity. *Proceedings of the National Academy of Sciences USA*, *113*, 674–679. https://doi.org/10.1073/pnas.1523683113

Rouhan, G., & Gaudeul, M. (2014). Plant taxonomy: A historical perspective, current challenges, and perspectives. In P. Besse (Ed.), *Molecular plant taxonomy. Methods in molecular biology (methods and protocols)* (Vol. *1115*). Totowa, NJ: Humana Press. https://doi.org/10.1007/978-1-62703-767-9_1

Sandel, B., Arge, L., Dalsgaard, B., Davies, R. G., Gaston, K. J., Sutherland, W. J., & Svenning, J. C. (2011). The influence of Late Quaternary climate-change velocity on species endemism. *Science*, *334*, 660–664. https://doi.org/10.1126/science.1210173

Scherrer, D., & Körner, C. (2011). Topographically controlled thermal-habitat differentiation buffers alpine plant diversity against climate warming. *Journal of Biogeography*, *38*, 406–416. https://doi.org/10.1111/j.1365-2699.2010.02407.x

Sklenář, P., Hedberg, I., & Cleef, A. M. (2014). Island biogeography of tropical alpine floras. *Journal of Biogeography*, *41*, 287–297. https://doi.org/10.1111/jbi.12212

Steinbauer, M. J., Field, R., Grytnes, J.-A., Trigas, P., Ah-Peng, C., Attorre, F., Birks, H. J. B., Borges, P. A. V., Cardoso, P., Chou, C.-H., De Sanctis, M., de Sequeira, M. M., Duarte, M. C., Elias, R. B., Fernández-Palacios, J. M., Gabriel, R., Gereau, R. E., Gillespie, R. G., Greimler, J., … Beierkuhnlein, C. (2016). Topography-driven isolation, speciation and a global increase of endemism with elevation. *Global Ecology and Biogeography*, *25*, 1097–1107. https://doi.org/10.1111/geb.12469

Steinbauer, M. J., Grytnes, J.-A., Jurasinski, G., Kulonen, A., Lenoir, J., Pauli, H., Rixen, C., Winkler, M., Bardy-Durchhalter, M., Barni, E., Bjorkman, A. D., Breiner, F. T., Burg, S., Czortek, P., Dawes, M. A., Delimat, A., Dullinger, S., Erschbamer, B., Felde, V. A., … Wipf, S. (2018). Accelerated increase in plant species richness on mountain summits is linked to warming. *Nature*, *556*, 231–234. https://doi.org/10.1038/s41586-018-0005-6

Testolin, R., Attorre, F., & Jiménez-Alfaro, B. (2020). Global distribution and bioclimatic characterization of alpine biomes. *Ecography*, *43*, 779–788. https://doi.org/10.1111/ecog.05012

Title, P. O., & Bemmels, J. B. (2018). ENVIREM: An expanded set of bioclimatic and topographic variables increases flexibility and improves performance of ecological niche modeling. *Ecography*, *41*, 291–307. https://doi.org/10.1111/ecog.02880

Visser, V., Clayton, W. D., Simpson, D. A., Freckleton, R. P., & Osborne, C. P. (2014). Mechanisms driving an unusual latitudinal diversity gradient for grasses. *Global Ecology and Biogeography*, *23*, 61–75. https://doi.org/10.1111/geb.12107

Vonlanthen, C. M., Kammer, P. M., Eugster, W., Bühler, A., & Veit, H. (2006). Alpine vascular plant species richness: The importance of daily maximum temperature and pH. *Plant Ecology*, *184*, 13–25. https://doi.org/10.1007/s11258-005-9048-5

Wade, L. K., & McVean, D. N. (1969). *Mt Wilhelm studies I. The alpine and subalpine vegetation*. Australian National University, Department of Biogeography.

Walter, H., & Box, E. (1976). Global classification of natural terrestrial ecosystems. *Vegetatio*, *32*, 75–81. https://doi.org/10.1007/BF02111901

Wang, Z., Brown, J. H., Tang, Z., & Fang, J. (2009). Temperature dependence, spatial scale, and tree species diversity in eastern Asia and North America. *Proceedings of the National Academy of Sciences USA*, *106*, 13388–13392. https://doi.org/10.1073/pnas.0905030106

Weigelt, P., Jetz, W., & Kreft, H. (2013). Bioclimatic and physical characterization of the world's islands. *Proceedings of the National Academy of Sciences USA*, *110*, 15307–15312. https://doi.org/10.1073/pnas.1306309110

Weigelt, P., Steinbauer, M. J., Cabral, J. S., & Kreft, H. (2016). Late Quaternary climate change shapes island biodiversity. *Nature*, *532*, 99–102. https://doi.org/10.1038/nature17443

Whittaker, R. J., Fernández-Palacios, J. M., Matthews, T. J., Borregaard, M. K., & Triantis, K. A. (2017). Island biogeography: Taking the long view of nature's laboratories. *Science*, *357*, eaam8326. https://doi.org/10.1126/science.aam8326

Whittaker, R. J., Triantis, K. A., & Ladle, R. J. (2008). A general dynamic theory of oceanic island biogeography. *Journal of Biogeography*, *35*, 977–994. https://doi.org/10.1111/j.1365-2699.2008.01892.x

Wiser, S. K. (2016). Achievements and challenges in the integration, reuse and synthesis of vegetation plot data. *Journal of Vegetation Science*, *27*, 868–879. https://doi.org/10.1111/jvs.12419

Wright, D. H. (1983). Species-energy theory: An extension of species-area theory. *Oikos*, *41*, 496–506. https://doi.org/10.2307/3544109

Xing, Y., & Ree, R. H. (2017). Uplift-driven diversification in the Hengduan Mountains, a temperate biodiversity hotspot. *Proceedings of the National Academy of Sciences USA*, *114*, E3444–E3451. https://doi.org/10.1073/pnas.1616063114

BIOSKETCH

Riccardo Testolin is a PhD candidate in environmental biology. He is mainly interested in macroecology, global change ecology, ecosystem modelling and ecoinformatics. His current research topic covers global alpine areas, with a focus on patterns and drivers of vascular plant diversity.

SUPPORTING INFORMATION

Additional Supporting Information may be found online in the Supporting Information section.

How to cite this article: Testolin R, Attorre F, Borchardt P, et al. Global patterns and drivers of alpine plant species richness. *Global Ecol Biogeogr*. 2021;30:1218–1231. https://doi.org/10.1111/geb.13297