1	Spatial patterns of species richness and nestedness in ant assemblages
2	along an elevational gradient in a Mediterranean mountain range
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20 Abstract

21 The study of biodiversity spatial patterns along ecological gradients can serve to 22 elucidate factors shaping biological community structure and predict ecosystem 23 responses to global change. Ant assemblages are particularly interesting as study cases. 24 because ant species play a key role in many ecosystem processes and have frequently 25 been identified as useful bioindicators. Here we analyzed the response of ant species 26 richness and assemblage composition to elevational gradients in Mediterranean 27 grasslands and subsequently tested whether these responses were stable spatially and 28 temporally. We sampled ant assemblages in two years (2014, 2015) in two mountain 29 ranges (Guadarrama, Serrota) in Central Spain, along an elevational gradient ranging 30 from 685 to 2390 m a.s.l.

31 Jackknife estimates of ant species richness ranged from three to 18.5 species and 32 exhibited a hump-shaped relationship with elevation that peaked at mid range values 33 (1100 - 1400 m). This pattern was transferable temporally and spatially. Elevation was 34 significantly related to ant assemblage composition and facilitated separation of higher 35 elevation assemblages (> 1700 m) from the remaining lower elevation species groups. 36 Ant assemblages were nested; therefore species assemblages with a decreased number 37 of species were a subset of the richer assemblages, although species turnover was more 38 important than pure nestedness in all surveys. The degree of nestedness changed non-39 linearly as a cubic polynomial with elevation. These assembly patterns were observed 40 over time but not between the two study regions.

We concluded double environmental stressors typical of Mediterranean mountains
explained species richness patterns: drought at low elevations and cold temperatures at
high elevations likely constrained richness at both extremes of elevational gradients.

44 The fact that species turnover showed a dominant role over pure nestedness suggested 45 current ant assemblages were context-dependent (spatio-temporal factors) and highly 46 vulnerable to global change, which threatens the conservation of present day native ant 47 communities, particularly at high elevations.

48

49 Introduction

50 Predicting the response of biodiversity to the main drivers of Global Change has 51 become a primary goal of modern ecology [1-3]. Consequently, the analysis of species 52 richness and assembly patterns along latitudinal and altitudinal gradients provide clues 53 to project possible effects of climate change on communities [4, 5]. The Mediterranean 54 mountains are particularly suitable for this purpose, given a characteristic combination 55 of temperature and water availability gradients in space and time, which determines a 56 progressively cooler and wetter environment at higher elevations while valleys suffer 57 more severe summer drought conditions [6, 7]. In addition, the Mediterranean Basin 58 will probable face particularly marked increases in aridity, temperature and frequency of 59 extreme climatic events [8-10]; therefore it is critical to advance our knowledge of how 60 basin communities respond to more severe abiotic factors.

Ants are diverse and ubiquitous, exhibit strong interspecific interactions, and perform many functional processes in ecosystems [11]; therefore ants are considered a viable indicator group of ecosystem change by many community ecologists. Ants are also dominant terrestrial community members [12] and play an important role in plant community dynamics, acting as seed harvesters [13], dispersal agents [14], and influencing soil nutrient status and plant growth [15]. As predators, ants control

herbivore abundance, drive relevant top-down effects, and are often used as biological
control agents of insect pests and fungal pathogens [16]. Furthermore, ant distribution
and abundance patterns provide information to address specific environmental
management issues [17, 18]; consequently, ants are particularly useful bioindicators.

71 Previous research on ant diversity documented variable responses to altitudinal 72 gradients, which showed a certain dependency on macroclimate. For instance, studies 73 reporting monotonic decreases in species richness, were frequent in temperate 74 mountains [19-21], while monotonic increases were observed in arid environments [22, 75 23]. Longino and Colwell [24] and Nowrouzi et al. [25] indicated more complex 76 patterns in tropical climates, which often combined a monotonic decrease at high 77 elevations with a plateau at low elevations. Mid-elevation peaks were reported in 78 tropical areas [26, 27], although also frequent at other latitudes [28, 29]. These patterns 79 were partially attributed to geometric constraints, such as the mid-domain effect and 80 area availability [30, 31] or the Rapoport rescue effect [32]. However the role of 81 climatic factors, including temperature [19-21] and water availability [22, 29, 33, 34] as 82 direct drivers of ant diversity is also unequivocal. Moreover, several studies 83 demonstrated these climatic variables also indirectly modulated ant diversity, for 84 example, by affecting primary productivity [33, 35-37].

Understanding how ant diversity responds to altitudinal gradients is not sufficient, however, to predict how environmental changes (e.g. rising temperatures) will affect species geographic ranges, because knowledge on species composition patterns is also required. Species assemblages and their distributions along ecological and geographical gradients result from species tolerances to climatic (abiotic) factors and biotic interactions, which interplay with niche evolution to determine community composition

91 [38-40]. Kodric-Brown and Brown [41] defined a structured community as one where 92 the organization is not due to randomness [41], and all situations in which communities 93 are not random can be explained by spatial turnover, nestedness, or combinations of 94 both [42]. Nestedness of species assemblages occurs when site biotas with smaller 95 species numbers are subsets of the biotas at richer sites [43, 44]. If environmental and 96 habitat filtering and not interspecific competition is responsible for nestedness, most 97 species can coexist. Therefore, under increasing temperatures, species could migrate or 98 expand their distribution elsewhere from the species former range, without necessarily implying other resident species will be displaced by the newcomers. On the contrary, 99 100 spatial turnover implies the replacement of some species by others, exhibiting gains and 101 losses of species from location to location globally [45]. This is a sign that some species 102 cannot coexist, because of different climatic requirements, competitive exclusion, or 103 both. In general, global warming will extend insect species ranges to higher altitudes 104 [46], therefore with spatial turnover patterns, species will migrate up an altitudinal 105 gradient and those at higher altitudes might suffer a subsequent reduction in geographic 106 distribution range, and even disappear from the area. However, Maguire et al. [47] 107 acknowledged both processes and community patterns might vary with scale, and 108 exhibit spatio-temporal variation, which raises concerns on our capacity to generalize 109 the results based on local studies. Indeed, transferability of results is increasingly 110 considered in analyzing the effects of global change on species distributions, although 111 such efforts are rarer in studies of community patterns [48-50].

In the present study, we explored patterns of ant species diversity and community composition in Mediterranean mountain grasslands, integrating spatio-temporal transferability. Specifically, we addressed the following questions: (1) what is the

relationship between ant species richness and elevation; (2) how does elevation affect ant species composition; (3) what are the relative contributions of spatial turnover and nestedness in ant community composition; and (4) are our findings transferable spatially and temporally?

119

120 Methods

121 Study areas and sampling design

122 We studied grassland ant communities in two mountains in the larger Sistema Central 123 range, Sierra de Guadarrama (reaching 2428 m a.s.l. at its highest peak) and Serrota (maximum 2294 m a.s.l.). The two mountains are ~ 100 km apart with similar climatic, 124 geologic, and biotic characteristics. Mean annual temperatures range from ~ 14 °C at 125 126 600 m to ~ 4 °C at the summits (2300 - 2428 m); and mean annual rainfall ranges from 127 550 mm to 1500 mm, with severe summer drought [51]. Substrata are primarily 128 composed of granites, and pasturelands are distributed along the complete elevational 129 gradient, largely as the result of traditional livestock grazing.

130 Sample sites were interspersed along the elevational gradient, facing southeast to 131 southwest with gentle slopes (< 5%), and not affected by anthropogenic disturbance 132 (i.e., not close to buildings or main paths). We sampled 18 grasslands in 2014 (in 133 Guadarrama range) and 12 in 2015 (six in Guadarrama range and six in Serrota range). 134 On each grassland, we randomly placed a 4 x 3 grid of pitfall traps, 2.5 cm diameter x 5 135 cm deep, with 5 m spacing between traps, which were filled with a 3:1 ethanol/ 136 monoethylene glycol mixture. Ant sampling was conducted in July, during stable 137 anticyclonic meteorological periods (Guadarrama: 17 - 23 July 2014 and 9 - 16 July

2015; Serrota: 1 - 8 July 2015), taking advantage of the seasonal ant activity peak
during summer in Central Spain [52]. Traps were collected after one week and ant
specimens were sorted to species, excluding winged ones.

Pitfall trapping is considered more objective and unbiased than other methods for sampling ground ants [53, 11]. We chose pitfall traps previously tested in other studies in Central Spain, which demonstrated success at capturing the complete species pool and the entire ant size range variability present in the sampled area [52, 54, 55].

145

146 <u>Statistical analyses</u>

We examined the relationship between species richness, taxonomic composition, and nestedness of grassland ant communities with elevation. Our analytical strategy was to build descriptive models for the data gathered during the most intensive sampling in Guadarrama 2014, and to validate the patterns detected with data gathered in a subset of the Guadarrama grasslands during 2015 (temporal validation) and with data gathered in the Serrota range during 2015 (spatial validation).

153 Species richness

First, we assessed the relationship between species richness and elevation. We estimated the species pool size at each locality with the incidence-based first order jackknife estimator. The relationship shape between estimates of richness and elevation was explored applying a generalized additive model (GAM) with Gaussian errors and a thin plate regression spline for elevation. Models were built with data from Guadarrama 2014 sampling and validated with data from the Guadarrama 2015 and Serrota 2015 samplings. Squared Pearson correlation between jackknife estimates of richness and 7

jackknife estimates predicted by the model were used to assess species richness patternconstancy between years and sites.

163 Taxonomic composition of assemblages

164 We explored how species assemblages differed based on taxonomic composition by 165 applying non-metric multidimensional scaling (NMDS) to the hemi-matrices of binary 166 Bray-Curtis distances among grasslands derived from species occurrences in pitfall 167 traps. In addition, a model-based analysis was conducted by building separate negative 168 binomial regressions for each species abundance data (as the multivariate response 169 variable) by elevation, year, and interaction (as the explanatory variables) [56]. 170 Significance of the interaction (assessed by resampling) would suggest that patterns of 171 species composition along an elevational gradient changed between years. Likewise, for 172 Guadarrama 2014 and Serrota 2015, a multivariate model was built by fitting negative 173 binomial regressions for individual species abundance, elevation, location, and 174 interaction [56]. Significance of the interaction would suggest how assemblages 175 changed with elevation were context-dependent.

176 Nestedness patterns

To assess assemblage nestedness among grasslands surveyed on each mountain and year we calculated the matrix temperature (T) [57] and the NODF (Nestedness metric based on Overlap and Decreasing Fill) [58]. While T has traditionally been the most commonly used metric for assessing overall nestedness [59], the more recent NODF index is claimed to exhibit more robust statistical properties, and separate contributions to nestedness of columns (due to species incidence) and rows (due to site composition) can be quantified [58, 60]. T decreases and NODF increases with nestedness.

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185 The significance of nestedness indices was estimated by comparison with suitable 186 binary null models, which were simulated by randomizing the original matrix 187 transformed to presence/absence binary data. Two null models were built to encompass 188 the range from maximally liberal (equitable) to maximally conservative (constrained) 189 null models [61]. The first was an equiprobable model built by randomizing rows and 190 columns and therefore maintaining only the number of species presences ('r00' 191 algorithm). The second was a proportional resampling model constructed by 192 constraining randomization to maintain row and column totals (site richness and species 193 incidence, respectively), while using marginal column frequency to select species 194 ('quasiswap' algorithm). Thus, the first null model was more liberal and only accounted 195 for matrix fill (the incidence of the total species set), while the second null model was 196 more stringent, because it imposed additional structure on the data, accounting for 197 among-site differences (e.g. different carrying capacities) and among-species 198 differences (e.g. different rarities). Subsequently, significant nestedness was attributed 199 to variation in observed species richness or species incidence, if evaluated with the 200 equiprobable model, or to variation beyond that observed, if evaluated with the 201 proportional resampling model [62]. One thousand randomizations of the original 202 matrix were applied.

The relative degree of nestedness, i.e. how much a grassland was nested within the set of sampled grasslands in the same geographic area and time period, was evaluated by a nestedness rank, according to T. Poorer grasslands, i.e. those having lower species diversity which were a subset of richer grasslands, were assigned higher ranks. These ranks were estimated as the ordinate in nestedness plots built with the T index, which

were calculated as (k - 0.5)/n for k = 1..., n rows (e.g. the bottom row in the graph, where k = 18 was the more nested site, which for the 2014 sampling in Guadarrama had a rank of (18 - 0.5)/18 = 0.97).

211 The shape of the relationship between nestedness and elevation during the 2014 212 sampling in Guadarrama (n = 18) was explored using GAM with Gaussian errors and a 213 thin plate regression spline for elevation. This model was validated using Guadarrama 214 2015 data (temporal validation). Detection of a high correlation between observed (in 215 2015) and predicted ranks would suggest the nestedness rank of plots did not vary 216 between years. Similarly, the model built with Guadarrama 2014 sampling data was 217 validated using the Serrota 2015 data (spatial validation). Detection of a high correlation 218 between observed and predicted ranks would suggest the nestedness rank of plots did 219 not vary between regions.

220 The contributions of spatial turnover and nestedness to the distribution pattern in 221 Guadarrama 2014 data were calculated using three beta diversity indices: Sørensen-222 based multiple-site dissimilarity (β_{SOR}), Simpson-based multiple-site dissimilarity 223 (β_{SIM}) , and nestedness-resultant multiple-site dissimilarity (β_{NES}) . β_{SIM} accounts for 224 spatial turnover and β_{NES} integrates dissimilarity due to nestedness, while β_{SOR} 225 expresses the total dissimilarity between communities and equals the sum of β_{SIM} and 226 β_{NES} [42]. *P*-values were estimated through the equiprobable ('r00') null model (the 227 proportional resampling null model was not useful to assess this partitioning of beta 228 diversity).

Statistical analyses were performed using R (v 3.3.2; R Core Team 2016) and
specialized packages vegan (v. 2.4-5) [63], mvabund (v. 3.12-3) [56], and betapart
(1.5.0) [64].
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233 Results

We detected 37 species, all recorded in Guadarrama (35 in 2014 and 26 in 2015, 24 were shared between years) and 20 species in Serrota (all species shared with Guadarrama). A total of 15 species were common among the three surveys (**S1 Table**).

237 Jackknife estimates of ant species richness in Guadarrama 2014 ranged from three (SE 238 = 0) species in most of the highest grasslands to 18.5 (SE = 2.6) at intermediate 239 elevations (Table 1). Estimates of richness related to elevation were positive below 240 1100 m a.s.l. and negative until a lower limit was reached at ca. 2000 m a.s.l. (Fig 1). 241 Overall, richness exhibited a hump-shaped relationship with elevation, which we 242 described using GAM with a spline for elevation, roughly equivalent to a third-order degree polynomial (equivalent df = 3.96, F = 12.94, P < 0.0001, $D^2 = 83.8\%$, Fig 1). 243 244 This pattern of species richness was similar between years and between mountain ranges (correlations for estimates of richness and richness predicted by the model were r 245 = 0.92, t = 4.68, df = 4, P = 0.0095, R² = 85% for Guadarrama 2015 data; and r = 0.86, t 246 = 3.42, df = 4, P = 0.0267, $R^2 = 75\%$ for Serrota 2015 data). 247

Table 1. Observed (S) and first-order jackknife estimator ($J \pm SE$) of ant species richness in dry grasslands in central Spain.

ID	Sampling	S	J±SE	N	Х	Y	Elevation (m)
G1	G2014	7	7±0	13	437876	4494480	685
G1	G2015	7	7.9±0.9	10	437876	4494480	685

			T				
G2	G2014	8	8±0	12	434821	4496549	768
S1	S2015	14	18.5±2.4	10	351711	4474792	808
G3	G2014	14	16.7±1.6	11	440394	4503515	865
G4	G2014	14	16.8±1.6	12	435541	4506094	1012
G5	G2014	14	15.8±1.3	12	431692	4513777	1044
G5	G2015	11	11.9±0.9	12	431692	4513777	1044
S2	\$2015	11	14.7±2.3	12	357653	4481333	1086
G6	G2014	13	14.8±1.3	9	414145	4507982	1171
G7	G2014	13	18.5±2.6	12	438808	4523233	1285
G8	G2014	15	15.9±0.9	11	416330	4511347	1330
G8	G2015	14	17.5±2.2	8	416330	4511347	1330
S3	S2015	13	15.8±1.6	12	340207	4476196	1374
G9	G2014	9	9±0	12	436806	4524537	1487
G10	G2014	10	12.8±1.6	12	400821	4495703	1625
S4	\$2015	4	4±0	11	326465	4482790	1646
G11	G2014	3	3±0	19	400543	4496999	1675
G12	G2014	5	5.9±0.9	12	419833	4519838	1786
G12	G2015	5	5.9±0.9	12	419833	4519838	1786
	l	1	1				

G13	G2014	5	5±0	12	419076	4519012	1830
G14	G2014	4	4.9±0.9	12	418737	4519177	1920
S5	S2015	2	2±0	12	323719	4482788	2000
G15	G2014	3	3±0	12	419392	4520829	2026
G15	G2015	3	3±0	8	419392	4520829	2026
G16	G2014	3	3.9±0.9	12	417883	4521249	2139
G17	G2014	4	5.8±1.9	12	418676	4521128	2266
S 6	S2015	3	3.9±0.9	12	323944	4485467	2286
G18	G2014	3	3±0	12	419420	4522544	2390
G18	G2015	3	3.9±0.9	12	419420	4522544	2390

Grassland plots were ordered following elevation (in m a.s.l.), identified with an ID, where G stands for Guadarrama, S stands for Serrota, and numbers represent grassland position in increasing order of elevation. Sampling: data from Guadarrama range in 2014 (G2014) and 2015 (G2015) and Serrota range in 2015 (S2015). N: samples sizes (number of pitfall traps recovered from each locality). Geographic coordinates (X and Y) were provided as UTM European Datum 1950.

- 256 Fig 1. Relationship between first-order jackknife estimates of ant species richness
- (see Table 1) and elevation in dry grasslands of a Mediterranean range in centralSpain.

Curve and 95% IC from GAM with Gaussian errors fit to the Guadarrama 2014 survey
(black circles). Richness estimates for the surveys in Guadarrama 2015 (grey circles)
and Serrota 2015 (open circles) were superimposed. Elevation values were slightly
jittered to avoid overlap among some points.

263

264 A NMDS solution was built that satisfactorily summarized the Bray-Curtis distances among grasslands in both years and study regions (stress = 0.10, linear R^2 between 265 266 Bray-Curtis and NMDS distances = 0.95). This scaling separated grasslands roughly according to elevation, but not regions or years and suggested an elevation threshold of 267 268 1700 m a.s.l. was a suitable elevation to distinguish the two groups (Fig 2). Multivariate GLMs for species abundances revealed significant elevation effects (P = 0.001) and a 269 270 significant region effect (P = 0.047), but significant effects were not detected for year or 271 any elevation interactions (Table 2). These results suggested taxonomic composition 272 varied mainly with elevation, but differences between regions cannot be excluded.

273

Fig 2. Non-metric multidimensional scaling (NMDS) of ant assemblages based on binary Bray-Curtis distances among grasslands calculated with species occurrences from pitfall traps in a Mediterranean range of central Spain.

Circles represent data from Guadarrama (black: year 2014, grey: year 2015) and Serrota
(open circles) ranges. Symbol sizes are proportional to the elevation of sampling plots.
As reference, circles are superimposed to a smooth elevation surface (estimated using a
GAM of elevation on a bivariate spline of NMDS scores). The positions of some
illustrative species are given as reference.

282

283 Table 2. Multivariate GLMs (negative binomial regressions) fitted to species abundances.

	Deviance	P values
a)		
Elevation	192.91	0.001
Year	50.27	0.649
Elevation x Year	5.99	0.997
b)		
Elevation	178.14	0.001
Region	84.44	0.047
Elevation x Region	18.85	0.618

284 *P*-values estimated by bootstrap (n = 1000). Significant *P*-values (P < 0.05) are highlighted in 285 bold. a) Guadarrama 2014 vs. Guadarrama 2015. b) Guadarrama 2014 vs. Serrota 2015.

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Ant species assemblages were nested, showing a pattern suggesting differences in species richness among localities were more important than differences in prevalence among species (nestedness among rows was higher than nestedness among columns, **Table 3, S2 Table**). Nestedness ranks for Guadarrama 2014 was described with GAM 291 and curvilinear term (spline) for elevation, roughly similar to a third-degree cubic polynomial (df = 3.89, F = 15.38, P < 0.0001, $D^2 = 85.6\%$, Fig 3). The subset of these 292 293 grasslands sampled in 2015 maintained their nestedness ranks and the model transferred to 2015 data (Pearson's correlation between predicted and observed ranks, r = 0.96, t =294 7.28, df = 4, P = 0.0018, $R^2 = 93\%$, S3 Fig). The model did not transfer between study 295 regions (r = 0. 74, t = 2.18, df = 4, P = 0.10, $R^2 = 54\%$, S4 Fig). Largely, nestedness 296 297 increased with elevation, with the exception of the lowest and highest elevation 298 grasslands.

Table 3. Nestedness indices for dry grassland ant assemblages in Guadarrama
range 2014 surveys (18 grasslands, 35 species).

NODF	NODFr	NODFc	Т	fill
38.31:	40.62:	37.72:	25.24:	
		0 001 0 100	0 001 0 010	0.23
0.001-0.015	0.001-0.001	0.001 -0.103	0.001 -0.912	

301

Indices were derived from the Nestedness metric based on Overlap and Decreasing Fill (NODF), NODFr for rows (i.e. localities) and NODFc for columns (i.e. species) and matrix temperature (T). Numbers under headings for indices provide the index values followed by *P*-values estimated by equiprobable ('r00') and proportional resampling ('quasiswap') binary null models. Significant *P*-values (P < 0.05) are in bold face. Fill: matrix fill (sum of 1s/sum of cells).

Fig 3. Generalized additive model (GAM) of nestedness ranks (based on nestedness temperature (T) index) for elevation in ant species assemblages in dry grasslands of a Mediterranean range in central Spain.

Curve and 95% IC from GAM with Gaussian errors fit to the Guadarrama 2014 survey
(black circles). Nestedness ranks for the Guadarrama 2015 (grey circles) and Serrota
2015 (open circles) surveys are superimposed. Low ranks indicate poorer species
community richness, where species are a subset of richer species communities.
Elevation values were slightly jittered to avoid overlap among some points.

317

318 Species turnover and nestedness were significant beta diversity components, although 319 species turnover among grasslands was more important than pure nestedness of 320 assemblages ($\beta_{SOR} = 0.88$, $\beta_{SIM} = 0.79$, $\beta_{NES} = 0.09$, all *P*-values = 0.001). This 321 relationship among beta diversity components did not vary among surveys (**S5 Table**).

322

323 Discussion

324 Ant species richness and taxonomic composition changed predictably with elevation in 325 the Mediterranean mountains we investigated, exhibiting largely consistent patterns 326 between years and among regions. These changes were more notable along mid-327 elevations and associated more clearly with differences of species richness among sites 328 than with prevalence or abundance among species. To our knowledge, this is the first 329 study specifically designed to describe ant community species richness and taxonomic 330 composition responses to elevational changes in the Mediterranean Basin. However, we 331 restricted our study to grasslands, therefore the results might not be a suitable model 17

332 system for other habitat types. Notwithstanding, grasslands are among the ecosystems
333 exhibiting greater economic and ecological relevance in the Mediterranean Basin [65,
334 66].

335 Species richness showed a hump-shaped relationship with elevation, peaking at ca. 1100 336 m a.s.l. This model depiction fit to the primary dataset (Guadarrama 2014), transferred 337 adequately to the 2015 Guadarrama dataset, and also the spatial validation dataset in the 338 2015 Serrota range. Despite geometric constraints as mid-domain effects were described 339 to explain these curve types [30], more recent studies tended to downplay the 340 importance of these mechanisms [24, 29, 67]. In fact, while mid elevation peaks in 341 species diversity were reported as relatively common [28, 29], these observations are far 342 from universal, suggesting regional factors play an important role in driving elevational 343 patterns of species diversity. Interestingly, research showed monotonic increases in ant 344 species richness were common in temperate climates [19-21], where elevation was 345 correlated with temperature. In arid climates, monotonic decreases were observed [22, 346 23], where hydric stress was ameliorated with altitude. Indeed, hydric stress and 347 temperature are the two main elevational correlates in Mediterranean climates, with 348 water availability limiting communities at low elevations, while at high elevations cold 349 temperature stress is more important [6, 7]. This double constraint might drive a hump-350 shaped spatial pattern of ant species richness in Mediterranean mountains.

Most studies on ant communities reported species richness positively responded to temperature [19-21, 24, 26, 27, 67, 68], but a number of findings also emphasized the role of water availability [22, 29, 33, 34, 69]. Cold temperatures can reduce local ant species richness by direct effects that limit ant activity, such as reduced foraging time [35, 70], and, at larger biogeographical scales, cold temperatures can also limit species 356 pools by lowering speciation rates [67]. Azcárate et al. [70] reported low humidity, in 357 turn, can limit colony activation by reducing foraging possibilities, and therefore 358 constitutes a main environmental filter. Alternatively, temperature and water availability 359 can facilitate ant species richness indirectly, these abiotic factors affecting primary 360 productivity and therefore the range and abundance of resources ant colonies exploit 361 [33, 35-37]. Diversity-productivity hypothesis states that as productivity increases, so 362 does the availability of energy and resources, so density and size of colonies also 363 increase. These relationships should increase local species richness by abundance, 364 lowering local extinction risks or activating a sampling mechanism [71, 72]. In 365 Mediterranean climates, primary productivity can be limited by cold temperatures and 366 summer drought [73], so maximum primary productivity values likely occur at 367 intermediate elevations, where summer drought is not too severe and temperatures are 368 not too cold. Therefore, if species richness increases with primary productivity, a 369 humped curve is also expected.

370 As anticipated, elevation also affected the taxonomic composition of ant communities. 371 We observed a ca. 1700 m threshold separating high mountain ant species communities 372 from those along the lower elevational gradient. This threshold was consistent among 373 regions and between years. Furthermore, although nestedness was a significant 374 component of beta diversity, most elevational shifts among communities were 375 attributable to species turnover. This result suggested the communities with lowest ant 376 species diversity (high mountain communities) were also the most unique in a regional 377 context. The singularity of high elevation ant communities was previously observed in 378 different biomes [22, 24, 34, 35], which reinforced the conservation value of the 379 Mediterranean mountains reported for other taxa [74, 75].

380 The examination of community nestedness suggested current ant assemblages are 381 context-dependent and likely highly vulnerable to global warming and other 382 anthropogenic changes. This is because, first, species turnover exhibited a dominant 383 role over pure nestedness, which indicated different and characteristic groups of ant 384 species occurred along an elevational gradient, and at the very least, ant species 385 community composition was distinct at high elevations (> 1700 m) (Fig 2). We want to 386 emphasize that if species distributions followed gradually changing abiotic conditions 387 (e.g. a decreasing temperature gradient with elevation), then we would expect a 388 concomitant change in the taxonomic composition of ant assemblages, rather than the 389 development of new, characteristic, species groups [42, 76]. It is also illuminating to 390 consider that nestedness arises with contributions from differences among species and 391 among sites [59]. Nestedness patterns elsewhere were explained as a simple sampling 392 mechanism, where the smaller species group in less diverse, i.e. poorer communities 393 were a subset of species in more diverse, i.e. richer communities, reportedly when the studied localities differed in resource availability (or productivity) and this factor 394 395 determined species richness [72, 77]. However, this implied species attributes notably 396 contributed to nestedness, because rarer species with small distributional ranges (often 397 specialists or poor-dispersers) were not generally sampled by the least diverse, i.e. 398 poorest communities, while the more common taxa exhibited a wide distributional 399 range [61]. In our study, the least diverse communities were more nested than diverse 400 ant communities, but nestedness among species was less important than among sites. 401 We conclude the sampling mechanism did not fully explain the nested patterns in the 402 ant communities we identified and habitat and abiotic filtering provided more plausible 403 explanations for the patterns. However, we carefully controlled by design the habitat 404 type (arid grasslands in gentle south facing slopes), which implied those habitat filters — 20

405 if an integral component - should be fine grained. Nestedness patterns along the 406 elevational gradient were temporally constant, but varied between mountain ranges, 407 indicating ant community composition was spatially dependent. Finally, an historic 408 explanation of current ant species distribution patterns in a landscape context is an 409 important element to include. We studied a mountain range within the Mediterranean 410 biome where major landscape changes have not occurred over the last thirty years. 411 Extensive afforestation was conducted during 1950's and a progressive abandonment of 412 the region took place between the 1960's and 1980's. Therefore, we assume all sites are 413 equally accessible from the regional species pool. However, we cannot disregard other 414 time-related mechanisms, which might explain community composition, such as priority 415 effects favoring one or more species, depending on their relative order of arrival to a site 416 [78], or metapopulation dynamics involving local extinctions and recolonizations in 417 periods beyond this two-year study [79].

418 Overall, species taxonomic composition varied non-linearly with elevation. The 419 insufficiency of the sampling mechanism emphasizes that ant assemblages are highly 420 vulnerable to global changes on climate and anthropogenic impacts, which might affect 421 species at the habitat and/or community level. It is unlikely species' range shifts that 422 track environmental change will maintain current ant species community composition.

423

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433 **References**

434 [1] Gilman SE, Urban MC, Tewksbury J, Gilchrist GW, Holt RD. A framework for
435 community interactions under climate change. Trends in Ecology & Evolution.
436 2010;25(6):325-331.

437 [2] Bellard C, Bertelsmeier C, Leadley P, Thuiller W, Courchamp F. Impacts of
438 climate change on the future of biodiversity. Ecology Letters. 2012;15(4):365-377.

439 [3] Oliver TH, Morecroft MD. Interactions between climate change and land use
440 change on biodiversity: attribution problems, risks, and opportunities. Wiley
441 Interdisciplinary Reviews: Climate Change. 2014;5(3):317-335.

[4] Wilson RJ, Gutiérrez D, Gutiérrez J, Montserrat VJ. An elevational shift in
butterfly species richness and composition accompanying recent climate change. Global
Change Biology. 2007;13:1873-1887.

[5] Devictor V, Julliard R, Couvet D, Jiguet F. Birds are tracking climate warming, but
not fast enough. Proceedings of the Royal Society B: Biological Sciences.
2008;275:2743-2748.

[6] Walter H, Breckle SW. Walter's vegetation of the earth: the ecological systems of
the geo-biosphere. 4th ed. Berlin. Springer; 2002.
22

450	[7] Schöb C, Armas C, Guler M, Prieto I, Pugnaire FI. Variability in functional traits
451	mediates plant interactions along stress gradients. Journal of Ecology. 2013;101(3):753-
452	762.

[8] Nogués-Bravo D, Araújo MB, Lasanta T, Moreno JIL. Climate change in
Mediterranean mountains during the 21st century. AMBIO: A Journal of the Human
Environment. 2008;37(4):280-285.

456 [9] Intergovernmental Panel on Climate Change (IPCC) 2013. Climate Change 2013:

457 The Physical Science Basis, Contribution of Working Group I to the Fifth Assessment

458 Report of the Intergovernmental Panel on Climate Change. In Stocker TF et al., editors.

459 Cambridge, UK. Cambridge University Press; 2013.

[10] Benavides R, Escudero A, Coll L, Ferrandis P, Ogaya R, Gouriveau F, Peñuelas J,
Valladares F. Recruitment patterns of four tree species along elevation gradients in
Mediterranean mountains: Not only climate matters. Forest Ecology and Management.
2016;360:287-296.

464 [11] Majer JD, Alonso LE, Schultz TR. 1st ed. Agosti D, Majer JD, Alonso LE,
465 Schultz TR, editors. Ants. Standard methods for measuring and monitoring biodiversity.
466 Washington D.C.: Smithsonian Institution Press; 2000.

467 [12] Cushman JH, Lawton JH, Manly BFJ. Latitudinal patterns in European ant
468 assemblages: variation in species richness and body size. Oecologia. 1993;95:30-37.

469 [13] Azcárate FM, Arqueros L, Sánchez AM, Peco B. Seed and fruit selection by

470 harvester ants, *Messor barbarus*, in Mediterranean grassland and scrubland. Functional

471 Ecology. 2005;19:273-283.

472 [14] Wolff A, Debussche M. Ants as seed dispersers in a Mediterranean old-field
473 succession. Oikos. 1999;84(3):443-452.

474 [15] Dean WRJ, Milton SJ, Klotz S. The role of ants nest-mounds in maintaining
475 small-scale patchiness in dry grasslands in Central Germany. Biodiversity and
476 Conservation. 1997;6:1293-1307.

477 [16] Philpott SM, Armbrecht I. Biodiversity in tropical agroforest and the ecological
478 role of ants and ant diversity in predatory function. Ecological Entomology.
479 2006;31(4):369-377.

480 [17] Underwood EC, Fisher BL. The role of ants in conservation monitoring: If, when481 and how. Biological Conservation. 2006;132(2):166-182.

[18] Schmidt FA, Ribas CR, Schoereder JH. How predictable is the response of ant
assemblages to natural forestry recovery? Implications for their use as bioindicators.
Ecological Indicators. 2013;24:158-166.

[19] Machac A, Janda M, Dunn RR, Sanders NJ. Elevational gradients in phylogenetic
structure of ant communities reveal the interplay of biotic and abiotic constraints on
diversity. Ecography. 2011;34:364-371.

[20] Reymond A, Purcell J, Cherix D, Guisan A, Pellissier L. Functional diversity
decreases with temperature in high elevation ant fauna. Ecological
entomology. 2013;38(4):364-373.

491 [21] Kwon TS, Kim SS, Chun JH. Pattern of ant diversity in Korea: An empirical test
492 of Rapoport's altitudinal rule. Journal of Asia-Pacific Entomology. 2014;17:161-167.

493	[22] Sanders NJ, Moss J, Wagner D. Patterns of ant species richness along elevational
494	gradients in an arid ecosystem. Global ecology and biogeography. 2003;12(2):93-102.

- 495 [23] Orabi GM, Semida FM, Abdel-Dayem MS, Sharaf MR, Zalat SM. Diversity
- 496 patterns of ants along an elevation gradient at St. Catherine Protectorate, South Sinai,
- 497 Egypt: (Hymenoptera: Formicidae). Zoology in the Middle East. 2011;54(1):101-112.
- 498 [24] Longino JT, Colwell RK. Density compensation, species composition, and 499 richness of ants on a neotropical elevational gradient. Ecosphere. 2011;2(3):1-20.
- 500 [25] Nowrouzi S, Andersen AN, Macfadyen S, Staunton KM, VanDerWal J, Robson
- 501 SKA. Ant Diversity and Distribution along Elevation Gradients in the Australian Wet
- 502 Tropics: The Importance of Seasonal Moisture Stability. PLoS ONE. 2016;11(4):
- 503 e0153420. doi:10.1371/journal.pone.0153420
- 504 [26] Munyai TC, Foord SH. Ants on a mountain: spatial, environmental and habitat 505 associations along an altitudinal transect in a centre of endemism. Journal of Insect 506 Conservation. 2012;16:677-695.
- 507 [27] Munyai TC, Foord SH. Temporal Patterns of Ant Diversity across a Mountain
 508 with Climatically Contrasting Aspects in the Tropics of Africa. PLoS ONE. 2015;10(3):
 509 e0122035. doi:10.1371/journal.pone.0122035
- 510 [28] Sanders NJ, Rahbek C. The patterns and causes of elevational diversity
 511 gradients. Ecography. 2012;35(1):1-3.
- 512 [29] Szewczyk T, McCain CM. A Systematic Review of Global Drivers of Ant
 513 Elevational Diversity. PLoS ONE. 2016;11(5): e0155404.
 514 doi:10.1371/journal.pone.0155404
 25

515	[30]	Colwell	RK,	Lees	DC.	The	mid-domain	effect:	geometric	constraints	on	the
-----	------	---------	-----	------	-----	-----	------------	---------	-----------	-------------	----	-----

516 geography of species richness. Trends in Ecology & Evolution. 2000;15(2):70-76.

517 [31] Sanders NJ. Elevational gradients in ant species richness: area, geometry, and

- 518 Rapoport's rule. Ecography. 2002;25(1):25-32.
- 519 [32] Stevens GC. The elevational gradient in altitudinal range: an extension of

520 Rapoport's latitudinal rule to altitude. The American Naturalist. 1992;140(6):893-911.

521 [33] Botes A, McGeoch MA, Robertson HG, van Niekerk A, Davids HP, Chown SL.

522 Ants, altitude and change in the northern Cape Floristic Region. Journal of523 Biogeography. 2006;33:71-90.

524 [34] Smith MA, Hallwachs W, Janzen DH. Diversity and phylogenetic community
525 structure of ants along a Costa Rican elevational gradient. Ecography. 2014;37(8):720526 731.

527 [35] Brühl CA, Mohamed M, Linsenmair KE. Altitudinal distribution of leaf litter ants
528 along a transect in primary forests on Mount Kinabalu, Sabah, Malaysia. Journal of
529 Tropical Ecology. 1999;15:265-277.

[36] Kaspari M, O'Donnell S, Kercher JR. Energy, Density, and Constraints to Species
Richness: Ant Assemblages along a Productivity Gradient. The American Naturalist.
2000;155(2):280-293.

533 [37] Gotelli NJ, Ellison AM. Biogeography at a regional scale: determinants of ant 534 species density in New England bogs and forests. Ecology. 2002;83(6):1604-1609.

- [38] Lavergne S, Mouquet N, Thuiller W, Ronce O. Biodiversity and climate change:
 integrating evolutionary and ecological responses of species and communities. Annual
 Review of Ecology, Evolution and Systematics. 2010;41: 321-350.
- [39] Laiolo P, Seoane J, Illera JC, Bastianelli G, Carrascal LM, Obeso JR. The
 evolutionary convergence of avian lifestyles and their constrained coevolution with
 species' ecological niche. Proceedings of the Royal Society B: Biological Sciences.
 2015;282(1821):20151808.
- 542 [40] Laiolo P, Seoane J, Obeso JR, Illera JC. Ecological divergence among young

543 lineages favours sympatry, but convergence among old ones allows coexistence in

544 syntopy. Global Ecology and Biogeography. 2017;26(5):601-608.

- 545 [41] Kodric-Brown A, Brown JH. Highly structured fish communities in Australian546 desert springs. Ecology. 1993;74:1847-1855.
- 547 [42] Baselga A. Partitioning the turnover and nestedness components of beta diversity.

548 Global Ecology and Biogeography. 2010;19:134-143.

- 549 [43] Wright DH, Reeves JH. On the meaning and measurement of nestedness of 550 species assemblages. Oecologia. 1992;92:416-428.
- 551 [44] Ulrich W, Gotelli NJ. Null model analysis of species nestedness patterns.
 552 Ecology. 2007;88(7):1824-1831.
- 553 [45] Gaston KJ, Davies RG, Orme CDL, Olson VA, Thomas GH, Ding TS,
- 854 Rasmussen PC, Lennon JJ, Bennett PM, Owens IPF, Blackburn TM. Spatial turnover in
- 555 the global avifauna. Proceedings of the Royal Society B: Biological Sciences.
- 556 2007;274:1567-1574.
 - 27

557 [46] Fielding CA, Whittaker JB, Butterfield JEL, Coulson JC. Predicting responses to
558 climate change: the effect of altitude and latitude on the phenology of the Spittlebug
559 *Neophilaenus lineatus*. Functional Ecology. 1999;13(1):65-73.

[47] Maguire KC, Nieto-Lugilde D, Fitzpatrick MC, Williams JW, Blois JL. Modeling
species and community responses to past, present and future episodes of climatic and
ecological change. Annual Review of Ecology, Evolution and Systematics.
2015;46:343-368.

564 [48] Kleinwächter M, Rickfelder T. Habitat models for a riparian carabid beetle: their

validity and applicability in the evaluation of river bank management. Biodiversity andConservation. 2007;16(11):3067-3081.

567 [49] Duncan RP, Cassey P, Blackburn TM. Do climatic envelop models transfer? A
568 manipulative test using dung beetle introductions. Proceedings of the Royal Society B:
569 Biological Sciences. 2009;267:1449-1457.

570 [50] Torres LG, Sutton PJH, Thompson DR, Delord K, Weimerskirch H, Sagar PM, et 571 al. Poor transferability of species distribution models for a pelagic predator, the grey 572 petrel, indicates constrasting habitat preferences across ocean basins. PLoS ONE.

573 2015;10(3): e0120014. doi:10.1371/journal.pone.0120014.

574 [51] i Casals MN, Pons X, i Nolla JMR. [Digital Climatic Atlas of the Iberian
575 Peninsula. Methodology and applications in bioclimatology and geobotany] Bellaterra
576 UAB; 2005. Spanish.

577 [52] Azcárate FM, Peco B. Abandonment of grazing in a mediterranean grassland area:
578 consequences for ant assemblages. Insect Conservation and Diversity. 2012;5:279-288.

579 [53] Andersen AN. Sampling communities of ground-foraging ants: Pitfall catches 580 compared with quadrat counts in an Australian tropical savanna. Australian Journal of 581 Ecology. 1991;16:273-279.

[54] Azcárate FM, Seoane J, Castro S, Peco B. Drove roads: Keystone structures that
promote ant diversity in Mediterranean forest landscapes. Acta Oecologica.
2013;49:107-115.

585 [55] Hevia V, Azcárate FM, Oteros-Rozas E, González JA. Exploring the role of 586 transhumance drove roads on the conservation of ant diversity in Mediterranean 587 agroecosystems. Biodiversity and Conservation. 2013;22:2567-2581.

588 [56] Wang YI, Naumann U, Wright ST, Warton DI. mvabund–an R package for 589 model-based analysis of multivariate abundance data. Methods in Ecology and 590 Evolution. 2012;3(3):471-474.

[57] Rodríguez-Gironés MA, Santamaría L. A new algorithm to calculate the
nestedness temperature of presence–absence matrices. Journal of Biogeography.
2006;33(5):924-935.

[58] Almeida-Neto M, Guimarães P, Guimarães Jr PR, Loyola RD, Ulrich W. A
consistent metric for nestedness analysis in ecological systems: reconciling concept and
measurement. Oikos. 2008;117:1227-1239.

597 [59] Ulrich W, Almeida-Neto M, Gotelli NJ. A consumer's guide to nestedness598 analysis. Oikos. 2009;118(1):3-17.

599 [60] Ulrich W, Gotelli NJ. Pattern detection in null model analysis. Oikos.600 2013;122(1):2-18.

- 601 [61] Nolte D, Schuldt A, Gossner MM, Ulrich W, Assmann T. Functional traits drive
- ground beetle community structures in Central European forests: Implications forconservation. Biological Conservation. 2017;213:5-12.
- 604 [62] Moore JE, Swihart RK. Toward ecologically explicit null models of 605 nestedness. Oecologia. 2007;152(4):763-777.
- 606 [63] Oksanen J, Blanchet FG, Friendly M, Kindt R, Legendre P, McGlinn D, et al.
- 607 Vegan: community ecology package R. Version 24-3 [software]. 2017 [cited 21 July

608 2018]. Available from: https://github.com/vegandevs/vegan.

- 609 [64] Baselga A, Orme CDL. betapart: an R package for the study of beta
 610 diversity. Methods in Ecology and Evolution. 2012;3(5):808-812.
- 611 [65] Joffre R, Rambal S, Ratte JP. The dehesa system of southern Spain and Portugal
 612 as a natural ecosystem mimic. Agroforestry Systems. 1999;45:57-79.
- [66] Bergmeier E, Petermann J, Schröder E. Geobotanical survey of wood-pasture
 habitats in Europe: diversity, threats and conservation. Biodiversity and Conservation.
 2010;19:2995-3014.
- [67] Sanders NJ, Lessard JP, Fitzpatrick MC, Dunn RR. Temperature, but not
 productivity or geometry, predicts elevational diversity gradients in ants across spatial
- 618 grains. Global Ecology and Biogeography. 2007;16(5):640-649.
- 619 [68] Malsch AKF, Fiala B, Maschwitz U, Mohamed M, Nais J, Linsenmair KE. An
- 620 analysis of declining ant species richness with increasing elevation at Mount Kinabalu,
- 621 Sabah, Borneo. Asian Myrmecology. 2008;2:33-49.
 - 30

[69] Jenkins CN, Sanders NJ, Andersen AN, Arnan X, Brühl CA, Cerda X, et al.
Global diversity in light of climate change: the case of ants. Diversity and Distributions.
2011;17:652-662.

[70] Azcárate FM, Kovacs E, Peco B. Microclimatic Conditions Regulate Surface
Activity in Harvester Ants *Messor barbarus*. Journal of Insect Behavior.
2007;20(3):315-329.

[71] Seoane J, Laiolo P, Obeso JR. Abundance leads to more species, particularly in
complex habitats: a test of the increased population size hypotheses in bird
communities. Journal of Biogeography. 2017;44:556-566.

[72] Storch D, Bohdalková E, Okie J. The more-individuals hypothesis revisited: the
role of community abundance in species richness regulation and the productivity–
diversity relationship. Ecology Letters. 2018;21:920-937.

[73] Flexas J, Diaz-Espejo A, Gago J, Gallé A, Galmés J, Gulías J, et al.
Photosynthetic limitations in Mediterranean plants: A review. Environmental and
Experimental Botany. 2014;103:12-23.

637 [74] Médail F, Quézel P. Biodiversity Hotspots in the Mediterranean Basin: Setting
638 Global Conservation Priorities. Conservation Biology. 1999;13(6):1510-1513.

639 [75] Sfenthourakis S, Legakis A. Hotspots of endemic terrestrial invertebrates in

southern Greece. Biodiversity & Conservation. 2001;10(8):1387-1417.

[76] Baselga A. Determinants of species richness, endemism and turnover in Europeanlonghorn beetles. Ecography. 2008;31:263–271.

643	[77] Evans KL, Warren PH, Gaston KJ. Species-energy relationships at the
644	macroecological scale: a review of the mechanisms. Biological Reviews. 2005;80:1-25.
645	[78] Wilbur HM, Alford RA. Priority effects in experimental pond communities:

responses of Hyla to Bufo and Rana. Ecology. 1985;66(4):1106-1114.

647 [79] Hanski I. Metapopulation dynamics. Nature. 1998;396:41-49.

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649 Supporting information

650 S1 Table. Species presence in each survey.

651 S2 Table. Nestedness indices for dry grassland ant assemblages in surveys from all

652 study areas in central Spain. Guadarrama range (-G2014- year 2014: 18 grasslands,

653 35 species; -G2015- year 2015: 6 grasslands, 26 species) and Serrota range (-S2015-

654 year 2015: 6 grasslands, 20 species). Indices are Nestedness metric based on Overlap

and Decreasing Fill (NODF), NODFr for rows (i.e., localities) and NODFc for columns

656 (i.e., species) and matrix temperature (T). Numbers under headings for indices provide

657 index values followed by *P*-values estimated by equiprobable ('r00') binary null models

and proportional resampling ('quasiswap'). Significant *P*-values (P < 0.05) are in bold

659 face. Fill: matrix fill (sum of 1s/sum of cells).

660 S3 Fig. Temporal validation of the relationship between nestedness and elevation.

661 Generalized additive model (GAM) of nestedness ranks (based on nestedness 662 temperature index) on elevation in ant assemblages from Guadarrama range (central

663 Spain) fit to Guadarrama 2014 data and validated with Guadarrama 2015 data.

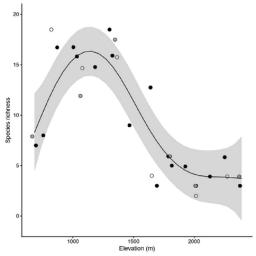
664 S4 Fig. Spatial validation of the relationship between nestedness and elevation.

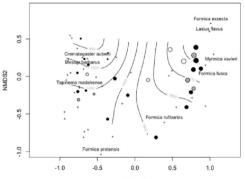
665 Generalized additive model (GAM) of nestedness ranks (based on nestedness 666 temperature index) on elevation in ant assemblages from Guadarrama range (central 667 Spain) fit to Guadarrama 2014 data and validated with Serrota 2015 data.

- 668 S5 Table. Multiple-site dissimilarities of beta diversity. Multiple-site dissimilarities
- accounting for the spatial turnover (β_{SIM}) and the nestedness components (β_{NES}) of beta
- 670 diversity, and sum of both values (β_{SOR}) for dry grassland ant assemblages in all study
- areas in central Spain. G2014: Guadarrama range 2014; G2015: Guadarrama range

672 2015; S2015: Serrota range 2015. P-values estimated by equiprobable binary null

- 673 models ('r00') are given between parentheses. Significant *P*-values (P < 0.05) are in
- bold face.





NMDS1

