

1 **Climate, and not fire, drives the phylogenetic clustering of species with**
2 **hard-coated seeds in Mediterranean Basin communities**

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21 **Abstract**

22 Hardseededness is a common trait in Mediterranean plant communities, although the origin of
23 its selection is controversial. It may be a mechanism of persistence to fire temperatures, but
24 could also form part of a gap-detecting mechanism to provide germination cues under arid
25 conditions. To disentangle this, we studied the phylogenetic structure of plant communities
26 against fire frequency and aridity gradients. The phylogenetic structure in Mediterranean Basin
27 ecosystems was analysed for the hardseededness trait as a whole and was separated by the
28 families composing this trait (Fabaceae and Cistaceae). This study focused on woody perennial
29 species. The phylogenetic structure was also contrasted against soil classes. Hardseededness
30 on the whole, and for the Fabaceae family alone, showed phylogenetic clustering as aridity
31 increased. Cistaceae displayed the opposite pattern with phylogenetic clustering in most
32 humid areas, together with a significant soil effect. Surprisingly, fire frequency had no
33 influence in any case. This climate-driven phylogenetic clustering indicates that the
34 hardseededness trait could confer some fitness advantage under dry conditions. For this
35 reason, coexisting species were more closely related in the community with increasing aridity.
36 This effect was especially evident for the Fabaceae family. These results shed some light on the
37 evolutionary selection of this adaptive trait under Mediterranean conditions. Our results
38 question the role of fire in the selection of the hardseededness trait in Mediterranean Basin
39 ecosystems and indicates that climate is the most important factor. Therefore, we should be
40 cautious in assigning to fire a preponderant role in the selection of some plant traits.

41 **Key words:** Adaptive trait; Community assembly; Habitat filtering; Hard coat; Seeder; Soils.

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45 1. Introduction

46 Habitat filtering is one of the most important processes that drives community assembly
47 (Webb et al., 2002; Kraft et al., 2014). Under a set of environmental conditions, specific traits
48 can confer advantages to some species *versus* others by allowing them to occupy a particular
49 habitat. For example, in Mediterranean ecosystems, climate and geology (soils) have been
50 traditionally considered the most fundamental forces of community assembly (Raven and
51 Axelrod, 1978; Thompson, 2005; Ackerly, 2009). However, it has been suggested in recent
52 years that fire can also be a factor of equal or greater importance in some cases (Verdú and
53 Pausas, 2007; Keeley et al., 2012). In this ecosystem type, fire is a recurrent disturbance caused
54 by hot dry summers that makes vegetation highly flammable (Keeley et al. 2012). Thus
55 community composition can be formed mainly by these three factors interacting with the pool
56 of species traits (Verdú and Pausas 2007, Ojeda et al. 2010).

57 Hardseededness (HS) is a frequent trait under Mediterranean conditions, but one that
58 is also present in many species and ecosystems worldwide (Parker and Kelly, 1989; Baskin et
59 al., 2000; Ooi et al., 2014; Willis et al., 2014; Rubio de Casas et al., 2017). In fire ecology
60 studies, HS is associated (or considered synonymous) with the physical dormancy conferred by
61 a hard-impermeable coat of palisade cells. This hard coat allows the formation of a persistent
62 soil seed bank that can last for decades (Parker and Kelly, 1989; Baskin et al., 2000). The
63 species with this trait are usually included in the so-called functional group of “seeders”,
64 characterised by a pulse of germination and establishment after disturbances (Paula and
65 Pausas, 2008). However, the origin of the selection of this adaptive trait is still controversial in
66 Mediterranean ecosystems (Bradshaw et al., 2011; Keeley et al., 2011; Santana et al., 2013a).
67 The selection of this trait has been long-since related to fire (Keeley, 1991; Moreira and
68 Pausas, 2012). Indeed the HS trait allows seeds to withstand high temperatures, while these
69 high temperatures break their impermeable coat and unleash a flush of germination at the
70 same time (Paula and Pausas, 2008). This flush occurs when resources availability is high by the

71 deletion of competing vegetation and the presence of a soil rich in nutrients through ash
72 deposition (Keeley, 1991). However, it has been suggested that HS trait selection could be
73 related to dry conditions, which are also typical of the Mediterranean climate (Baeza and Roy,
74 2008; Santana et al., 2013a). In the bare soil gaps created by disturbances, daily soil
75 temperature fluctuations in summer can break the physical dormancy of soil-stored seeds
76 (Baeza and Roy, 2008; Santana et al., 2013a; Ooi et al., 2014). This gap-detecting mechanism
77 may be an adaptive advantage because it coordinates germination after summer, during the
78 next wet autumn season, with enough time for seedlings to grow and establish before the next
79 dry summer season (Santana et al., 2013a; Jaganathan, 2015). One key attribute suggested to
80 distinguish between fire-adapted from gap-detecting species is the temperature threshold
81 required to break seed dormancy (Moreira and Pausas, 2012; Santana et al., 2013a; Ooi et al.,
82 2014). In species with obligate fire dormancy release, this threshold is closely related to
83 temperatures that occur only during fires (80-100°C), while in species with gap dormancy
84 release would be related to temperatures typical from summer soil temperatures (40-70°C)
85 (Ooi et al., 2014). However, very little information is available about this temperature
86 threshold, and further approaches are needed to clarify the role of different environmental
87 factors in HS trait selection.

88 In the Mediterranean Basin (MB), the HS trait is a phylogenetic conserved trait that is
89 present mainly in species from the Fabaceae and Cistaceae families (Verdú and Pausas, 2007).
90 In this region, these families have been detected to date back to the Neogene period, but their
91 presence intensified in the Quaternary, concomitantly with the definitive onset of a dry
92 seasonal climate (Mediterranean climate) and frequent fires (Thompson, 2005; Rundel et al.,
93 2016). The synchronised emergence of such factors does not clarify which of them was the
94 most determinant driver of community assembly processes, and whether HS selection actually
95 favoured the species with this trait. If environmental filtering was a dominant community
96 assembly process, then the HS trait would be expected to be clustered (Webb et al., 2002);

97 that is, coexisting species would share a niche trait that confer them some advantage. In
98 contrast, when environmental filters are absent, the community is shaped by competitive
99 interactions and is characterised by phylogenetic overdispersion (Webb et al., 2002; Cavender-
100 Bares et al., 2004). In the latter case, coexisting species are not closely related because they
101 tend to avoid niche overlapping. According to these premises, the assessment of which
102 environmental filter is the most determinant in clustering the HS trait, whether it is fire or
103 climate, may shed some light on its adaptive selection. In other words, if HS is an important
104 trait to be able to persist high fire activities or dry conditions, phylogenetic clustering would be
105 expected to increase in gradients of these factors. Some weaknesses have been pointed out in
106 using community phylogenetics to describe community assembly processes (Kraft et al., 2014;
107 Gerhold et al., 2015), but these methods are still very valuable for addressing eco-evolutionary
108 questions (Gerhold et al., 2015). In any case, this approach is not so simplistic and the major
109 drawback for determining which factor (fire or climate) is the most determinant is that they
110 are often conflated. On local and regional scales, both gradients often follow the same pattern
111 with higher fire frequencies at dry and hot sites. For example, phenotypic and phylogenetic
112 clustering of seeder species has been observed in the MB when comparing wet areas with low
113 fire frequency to drier areas with higher fire frequency (Lloret et al., 2005; Verdú and Pausas,
114 2007). Nevertheless, if we are to know the relative importance of fire or climate on HS trait
115 selection, it is essential that we distinguish between these two environmental gradients.

116 One plausible alternative to discern between fire activity and climatic gradients is the
117 intermediate fire-productivity hypothesis (Van der Werf et al., 2008; Krawchuk and Moritz,
118 2011). This hypothesis suggests that, with an increasing gradient of aridity/productivity, fire
119 activity peaks at intermediate levels by following a humped or unimodal relation. Here fuel
120 and drought act as fire drivers along the aridity/productivity gradient; i.e., in moist-productive
121 regions, fire activity is driven by frequency of droughts because fuel is readily available, while
122 fire regimes are fuel-limited in unproductive-arid systems. This hypothesis has been tested and

123 corroborated on different scales, ranging from regional (Pausas and Bradstock, 2007) to
124 worldwide gradients (Van der Werf et al., 2008; Krawchuk and Moritz, 2011; Murphy et al.,
125 2011), and also in MB ecosystems (Pausas and Paula, 2012). This hypothesis, therefore,
126 presents an ideal framework to analyse the community assembly processes that derive from
127 the environmental filtering of the HS trait by fire or climate.

128 It is noteworthy that fire and climate are not the only environmental factors to
129 influence the phylogenetic structure of plant communities in Mediterranean ecosystems. Soils
130 can also play an important role in these processes because they influence resource availability
131 (Dantas et al., 2015). In soils that are infertile or possess a low water retention capacity, the
132 traits that enhance seedling survival may also be fundamentally valuable. Indeed the HS trait
133 can imply an advantage through the conferred bet-hedging strategy (Venable, 2007).
134 Persistent soil seed banks, which annually release a small proportion of seeds to germinate,
135 spread the risk of seedling establishment failure over time (Fenner and Thompson, 2005).
136 Indeed it has already been observed that infertile soils can locally promote the phylogenetic
137 clustering of the “seeder” trait in MB ecosystems either by themselves or via an interaction
138 with fire (Ojeda et al. 2010). Therefore, studies that determine the phylogenetic structure of
139 plant communities should not neglect the influence of soil on community assembly processes.

140 The aim of this paper is to assess the role of fire, climate and soils on filtering the
141 presence of the HS trait in MB ecosystems. Our experimental design allowed us to define two
142 different and contrasting gradients of aridity and fire in the SE Iberian Peninsula: (i) a climatic
143 gradient of increasing temperatures and decreasing precipitation; (ii) a unimodal fire activity
144 gradient where higher frequencies occur at intermediate climatic gradient values (following
145 the fire-productivity hypothesis). We hypothesised that if fire predominated as the filtering
146 force, the phylogenetic clustering of the HS trait would follow a unimodal pattern on the
147 aridity gradient. In contrast, if climate was the most determinant force, phylogenetic clustering
148 would increase with aridity. We also included the relation between the phylogenetic structure

149 of communities and soil properties in this analysis. If it was a determinant, phylogenetic
150 clustering would increase with both soil infertility and lower water retention capacity.
151 Disentangling which factor plays a more determinant role in the phylogenetic clustering of the
152 HS trait could shed some light on the evolutionary selection of this trait. The hypotheses
153 herein proposed were tested for the HS trait on the whole, and were separated by the families
154 composing this trait (Fabaceae and Cistaceae). We intended to see if a differential response
155 corresponded to the intrinsic characteristics of the studied families.

156 **2. Material and Methods**

157 *2.1. Study area and vegetation data*

158 This work centred on the Valencia Region (Eastern Iberian Peninsula; 22885 km²; Figure 1),
159 which has a typical Mediterranean climate where forest fires are recurrent episodes. To study
160 the phylogenetic structure of the HS trait at the community level, we obtained the data
161 collected by the “Banc de dades de la Biodiversitat de la Comunitat Valenciana”
162 (<http://bdb.cma.gva.es>, last accessed in March 2014). This is a biodiversity bank data managed
163 by the Regional Valencian Government which, for each 1×1 UTM coordinate, collects all the
164 plant and animal species historically cited for this point. This bank has more than 19,000
165 species and 1,750,000 species detections throughout the region. This makes it an excellent
166 tool to assess the presence of species at the spatial level and for macro-ecological studies. For
167 this study, all the plant species present in the bank data were named and classified as life
168 forms (following Raunkiaer, 1934) according to the flora described for the whole Valencia
169 Region by Mateo and Crespo (2001). Only woody perennial biological types (i.e.,
170 chamaephytes and phanerophytes) were used. Species from coastal ecosystems or those
171 typical of specific ecosystems, such as riverlands and salt marshes, were removed, along with
172 cultivated, naturalised and exotic species. As the dataset is composed of individual
173 observations of species for each 1×1 UTM coordinate, thus they do not have equal sampling

174 efforts, we used the species presence-absence data matrix for 10×10 UTM. This smaller work
175 scale resulted in a reduction in the species-absence bias and ensured a representative dataset
176 of species presence. We took the sum of all the species present in this coordinate as the
177 corresponding community for each 10×10 UTM. Then we confirmed that each coordinate had
178 representative descriptions of the community composition by calculating the species
179 accumulation curve. We assumed that all the coordinates contained a representative
180 community composition as they achieved an equilibrium with an asymptote. The UTM
181 coordinates that did not meet these assumptions were removed from the analysis. These
182 coordinates were mainly those on the border limits of the region or in coastal areas with small
183 areas to be sampled, as well as the coordinates for which most of their area had been
184 urbanised (i.e., city of Valencia and conurbations). Finally, we used 202 UTM coordinates for
185 our study.

186 2.2. *Climatic gradient*

187 The climatic conditions for each 10×10 UTM coordinate were defined by including the values
188 of the mean annual precipitation (PP) and the mean annual temperature (T) in the database
189 (Pérez-Cueva, 1994). The whole of the Valencia Region is characterised by having a wide range
190 of aridity conditions. For example, the mean annual precipitation ranges from 817 mm to 217
191 mm, and comprises semiarid, dry subhumid and humid Mediterranean climates (Trabucco and
192 Zomer, 2009). The mean annual temperature also considerably varies from 10°C to 19°C. As a
193 climatic proxy of the aridity conditions, we used the Rainfall Index of Lang (1915) as the
194 quotient PP/T , which is considered one of the simplest and most direct ways of describing this
195 variable. The climatic gradient defined for the Valencia Region roughly follows a North to
196 South pattern, with some variability as a result of mountain ranges and its proximity to the sea
197 (Figure 1). Altitude ranges from 1,813 m to sea level.

198 2.3. *Fire gradient*

199 The fire activity in the Valencia Region follows a humped pattern along the defined aridity
200 gradient (Figure 2). The fire gradient follows the intermediate fire-productivity hypothesis,
201 with low fire activity in semiarid areas and humid areas as a result of biomass limitation and
202 the low frequency of fire-prone dry conditions, respectively (Pausas and Paula, 2012). Fire
203 activity concentrates mainly in intermediate areas, where a dry subhumid climate dominates.
204 We defined this humped pattern by two different proxies: number of fires (Nfires) and total
205 area burned (Burnt_area) since 1993. These data were provided by the regional environmental
206 services for each 10×10 UTM coordinate, and the two proxies were compared to the above-
207 described PP/T climatic index (Figure 2). Data included both natural and man-caused fires.

208 *2.4. Soil properties*

209 For each 10×10 UTM, a percentage of the area was occupied by different lithology classes
210 (COPUT, 1998). Therefore as a determinant of soil properties, we used these lithology classes
211 categorised into 11 different types: Limestones (36% of total area), Gravels (21%), Dolomites
212 (11%), Limes (8%), Marls (8%), Clays (5%), Conglomerates (5%), Sands (4%), Sandstones (3%),
213 Metamorphic (0.3%), Volcanic (0.01%). In order to include the soil differences in each 10×10
214 UTM, we performed a principal components analysis (PCA) with standardised data values. The
215 site scores of the first and second PCA axes were used as proxies of soil variability. The first
216 axis explained 15% of soil variability, while the second axis explained 13% (Figure S1 in the
217 Supplementary Material). PCA 1 showed a gradient of soils dominated by limes towards the
218 soils dominated by clays (left to right), with intermediate values dominated by Sands, Gravels,
219 Marls and Limestones (Soil_PCA1 hereafter). The PCA 2 axis showed a pattern that went from
220 positive values dominated by Limestones, Marls and Conglomerates to negative values
221 dominated by Sandstones, Sands and Gravels (Soil_PCA2 hereafter).

222 *2.5. Data analysis*

223 All the statistical analyses were performed in the R statistical environment (version 3.3.2 R
224 Core Team, 2016). Phylogenetic analyses were done using the packages *brranching*
225 (Chamberlain, 2016), *ape* (Paradis et al., 2004), *phytools* (Revell, 2012) and *picante* (Kembel et
226 al., 2010). The tree was visualised using *ggtree* (Yu et al., 2017). The phylogenetic analyses of
227 the DNA data were conducted on the CIPRES Science Gateway (Miller et al., 2010) and
228 processed with R packages. The multivariate analysis was performed with the *vegan* package
229 (Oksanen et al., 2017), while the generalised additive models were fitted using the *mgcv*
230 package (GAM; Wood, 2011).

231 2.6. Phylogenetic analysis

232 For the whole presence-absence data matrix (432 species), we assigned a qualitative state of
233 the HS trait (yes/no) to each species. In the MB, this trait is restricted to species from the
234 Cistaceae and Fabaceae families in woody perennial species (Baskin and Baskin, 2014). As we
235 were unable to ensure the proper assignment of the HS trait to herbaceous species, we
236 focused our study on woody perennial species. Not in vain, woody perennial species are the
237 main component of the composition and structure of the studied Mediterranean vegetation.
238 We had 90 species with the HS trait, 35 from Cistaceae and 55 from Fabaceae.

239 Afterwards, a phylogenetic tree was assembled for the 432 species included in the
240 dataset (Figure S2). For this purpose, we used the *phylomatic* function from the *brranching*
241 package. We constructed a tree by grafting families, genera and species onto the angiosperm
242 tree R20120829 for plants (Chamberlain, 2016). Once the tree was assembled, it was checked
243 for polytomies and resolved whenever possible. Branch lengths were adjusted by Grafen's
244 method (Malhado et al., 2015), and then it was forced to be ultrametric (function
245 *force.ultrametric* from the *phytools* package).

246 Secondly, we used DNA sequences to create a more resolved tree to reduce some of
247 the unresolved polytomies of the previous tree. Here data mining was done in the public

248 database GenBank to search for relevant sequences. Only one sequence per species was
249 downstreamed, and we were able to collect 245 sequences containing the 18S, 5.8S genes, the
250 complete ITS marker and of similar length. The multiple sequence alignment algorithm MAFFT
251 (Kato et al., 2005), available on the EMBL-EBI portal (<http://www.ebi.ac.uk/Tools/msa/mafft/>,
252 last accessed December 2019), was used to align the 245 sequences. The Maximum Likelihood
253 analyses were conducted using the RAxML HPC2 (Stamatakis 2006) in the XSEDE tool
254 (Phylogenetic tree inference using maximum likelihood/rapid bootstrapping run on XSEDE)
255 employing a GTR + G+ I substitution model. To assess the statistical support of clades, 1,000
256 fast-bootstrap (BS) replications were run according to the GTR-CAT approximation. However,
257 we were unable to retrieve the sequences for all the species (i.e., 247 species of 432; 57% of all
258 species). Here 110 species were added to their genus over this tree using the species genus by
259 defining a function based on the *add.species.to.genus* function of the phytools package (Figure
260 S3). Finally, we decided to utilise this phylogenetic tree based on DNA to resolve the
261 polytomies of some species from a previously built complete phylogenetic tree, and use it as
262 the working tree (Figure S2).

263 Initially, the phylogenetic structure of communities was assessed by calculating the
264 mean pairwise distance (MPD) of species with the HS trait included in each 10×10 UTM against
265 the MPD values obtained with a null model. Here the *ses.mpd* function (*picante* package) for
266 the richness data with a null-model of *taxa.labels* was used. This method reshuffles the species
267 labels across the above-assembled phylogenetic tree. Finally, in order to account for the
268 possible effect of different sampling efforts on species richness among sites, we derived the
269 net relatedness index (NRI) for the species with the HS trait, which is the standardised effect
270 size of MPD vs. null MPD (Webb et al., 2005). The NRI standardises the differences between
271 the average phylogenetic distances in the observed and null communities by the standard
272 deviation of the phylogenetic distances in null communities (Webb et al., 2005). The NRI rises
273 with increasing clustering and becomes negative with overdispersion (Webb et al., 2005). The

274 NRI evaluation of the Cistaceae and Fabaceae families was made by repeating the analysis, but
275 by considering only the species of each family.

276 2.7. *Phylogenetic structure against the climate, fire and soil gradients.*

277 To test our hypothesis about the main drivers of phylogenetic clustering of the HS trait, we
278 used GAM models. We examined how the HS trait, and the Cistaceae and Fabaceae NRI, are
279 related to the climate (PP/T index), soil (Soil_PCA1, Soil_PCA2) and fire variables (Nfires and
280 Burnt_area). This approach was chosen because it makes no *a priori* assumption about the
281 functional relation between variables (Wood, 2011), and depicts the real trend with no
282 restrictions. We considered that the phylogenetic structure was climate-driven if HS clustering
283 increased linearly with the climatic gradient. However, if it was driven by fire, phylogenetic
284 clustering would follow a hump-shaped pattern. Similarly, if soil was a determinant factor, the
285 phylogenetic structure would vary along the PCA soil gradients.

286 For each NRI (HS trait, Cistaceae and Fabaceae), we started fitting the null GAM model,
287 and then all the explanatory variables were added sequentially until the fully parameterised
288 model was fitted. Thin plate regression splines with shrinkage terms were used as the basis for
289 smooth climate, fire and soil explanatory variables. We simultaneously accounted for spatial
290 autocorrelation by including the coordinates of points (i.e., Latitude and Longitude) as smooth
291 interaction terms (Wood, 2011). Prior to GAM modelling, we eliminated the outliers and
292 extreme observations to reduce leverage using Cook's distance (Alimadad et al., 2011). We
293 then ranked models according to their Akaike's information criterion (AIC) and Bayesian
294 information criterion (BIC). The models with lower AIC and BIC values and the fewest
295 estimable parameters were considered the "best models". Here complex models were
296 considered a candidate model only if they had two AIC units less than the AIC of all their
297 simpler nested models (Burnham and Anderson, 2002). We simultaneously tested that the
298 variance inflation factor of the selected models was below 3, which indicates lack of

299 collinearity issues (Dormann et al., 2013). We used likelihood ratio R^2 and deviance proportion
300 as measures of the variance that the model explained.

301 Finally, in order to know the weight of the explanatory variables in determining the
302 NRIs (HS trait, Cistaceae and Fabaceae), all the variables were classified into four general
303 components of influence: climate variables (PP/T), soil variables (Soil_PCA1, Soil_PCA2), fire
304 variables (Nfires and Burnt_area) and spatial variables (Latitude, Longitude). Then two full
305 models of three components were defined: a first one with the climate and fire variables as
306 the main drivers, and a second model with climate and soil:

307 1) $y_i = a + s_1(\text{Latitude, Longitude}) + s_2(\text{PP/T}) + s_3(\text{Burned_area}) + s_4(\text{Nfires})$

308 2) $y_i = a + s_1(\text{Latitude, Longitude}) + s_2(\text{PP/T}) + s_3(\text{Soil_PCA1}) + s_4(\text{Soil_PCA2})$

309 Thus partial deviances were calculated sequentially by removing the predictors
310 corresponding to space, fire, climate and soil from these models. These partial deviances were
311 a proportion of the total model deviance explained by each component (Wood, 2011). Models
312 were refitted after the sequential elimination of components, in which two full models of
313 three components each were evaluated instead of a full model with the four components to
314 reduce the number of sequential possibilities.

315 **3. Results**

316 *3.1. HS trait*

317 Using generalised additive models, we were able to select the best NRI model for the HS trait
318 based on AIC, BIC and the fewest number of estimable parameters (df). This model included
319 the space (Latitude, Longitude) and climate (PP/T), but not the fire or soil variables (Table S1a).
320 The selected model explained a large proportion of deviance (40%; $R^2 = 0.35$), and the PP/T
321 predictor was significant (edf=5.39, $F=6.50$, $p < 0.001$). The NRI showed a linear decrease from
322 the positive clustered sites at lower PP/T values to the negative overdispersed sites, which

323 were stable around -0.5 with medium and high PP/T values (Figure 3). This meant that HS
324 clustering increased as aridity rose. Surprisingly, the fire variables were poorly related to the
325 NRI, did not appear in the best models and showed no humped pattern as predicted.

326 We divided predictors into groups of variables to represent space (Latitude,
327 Longitude), climate (PP/T), soil (Soil_PCA1 and Soil_PCA2), and fire (Nfires and Burnt_area).
328 We followed a deviance partitioning approach to identify the unique contributions of each
329 general component to the HS trait NRI. The results showed that space explained 27-28% of the
330 total explained deviance, and climate, soil and fire explained approximately 15-16%, about 4%,
331 and only 3%, respectively (Table 1a). These values indicated that the climatic gradient was the
332 main environmental variable that explained HS trait phylogenetic clustering variability despite
333 the strong influence of the spatial gradient.

334 3.2. *Fabaceae and Cistaceae responses*

335 The best NRI model for Fabaceae also included the space (Latitude, Longitude) and climate
336 (PP/T), but not the soil or fire variables (Table S1b). This model explained a large proportion of
337 deviance (49%; $R^2 = 0.45$), and all the included predictors were significant: space (edf=7.94,
338 $F=2.13$, $p=0.019$) and climate (edf=6.88, $F=7.55$, $p<0.001$). The NRI showed a similar response
339 to the HS trait, with a linear decrease from the positive clustered sites at lower PP/T values to
340 the negative overdispersed sites at higher PP/T values (Figure 3). For Fabaceae, when the
341 predictors were divided into groups of the variables representing space (Latitude, Longitude),
342 climate (PP/T index), soil (Soil_PCA1 and Soil_PCA2) and fire (Nfires and burnt_area), the
343 results once again showed that the spatial (24-26%) and climate (20-23%) patterns mainly
344 explained phylogenetic clustering, followed at some distance by the soil (6%) and fire (3%)
345 variables (Table 1b).

346 The Cistaceae NRI best model based on AIC, BIC and the fewest estimable parameters
347 (edf) included space (Latitude, Longitude), climate (PP/T) and soil (Soil_PCA1) (Table S1c). The

348 model explained a large proportion of deviance (69%, $R^2 = 0.64$), and all the included predictors
349 were significant: space (edf=19.90, $F=4.95$, $p<0.001$), climate (edf=4.40, $F=2.60$, $p<0.022$) and
350 soil (edf=1, $F=11.06$, $p=0.001$). Here the inclusion of fire did not considerably improve either
351 the AIC or BIC in the model (Table S1c). The NRI was clustered at the sites with higher PP/T
352 values, and lowered and remained overdispersed with values below 40 PP/T (Figure 3).
353 Regarding soils, the NRI clustered in the more negative Soil_PCA1 values and moved linearly to
354 overdispersion towards the positive values (Figure S4). It showed a spatial distribution for the
355 clustered NRI values (>0) located in the northern part of the territory, and lowered towards the
356 central and southern regions where the Cistaceae NRI was overdispersed (<0) (Figure 4). For
357 Cistaceae, when predictors were divided into groups of variables representing space (Latitude,
358 Longitude), climate (PP/T), soil (soil_PCA1 and soil_PCA2) and fire (Nfires and Burnt_area), the
359 results revealed that the spatial (32-39%) and climate (20-27%) patterns chiefly explained
360 Cistaceae clustering, followed by the soil variables at some distance (17%) and then by the fire
361 variables (2%) (Table 1c).

362 **4. Discussion**

363 Phylogenetic clustering did not follow the unimodal gradient followed by fire in our study area
364 in any case, which questions its role in HS trait selection in MB ecosystems (Verdú and Pausas,
365 2007; Moreira and Pausas, 2012). Our results better agree with the classic point of view that
366 climate and soils are the main drivers of community assembly processes in Mediterranean
367 ecosystems (Raven and Axelrod, 1978; Thompson, 2005; Ackerly, 2009; Keeley et al., 2012).
368 These results support the view that we should be cautious to assign a preponderant role to fire
369 in the selection of plant traits in some cases (Bradshaw et al., 2011). Of all the study variables,
370 climate was the main driver of the phylogenetic structure for the HS trait in MB communities.
371 Phylogenetic clustering for the whole HS trait followed an increasing pattern as aridity rose,
372 which was especially evident for the Fabaceae family. Conversely, the Cistaceae family

373 followed a different pattern by showing phylogenetic clustering in the most humid areas, as
374 well as a significant soil type effect. The spatial component also explained a large proportion of
375 HS trait variability as a result of the similarity expected between neighbouring communities
376 (spatial autocorrelation) and other factors not herein considered, such as dispersal ability,
377 species extinction or historic land use.

378 HS is inherent for species from the Fabaceae and Cistaceae families in the MB. The
379 presence of these families in this area was first observed during the Neogene (Thompson,
380 2005). The *Genista* and *Astragalus* genera (Fabaceae) appear in Pliocene records (Pignatti,
381 1978; Thompson, 2005) when the climate seasonality onset of the Mediterranean climate
382 intensified (Suc, 1984). Similarly, the molecular clock places the origin of the *Fumana* and
383 *Helianthemum* genera (Cistaceae) during the Miocene (Guzmán and Vargas, 2009).
384 Palynological records of Cistaceae species have been dated for the Miocene, and were
385 probably hosted in semiarid open vegetation spots when the tropical climate began to
386 undergo seasonal variation (Thompson, 2005; Jiménez-Moreno, et al. 2010). It was during this
387 transition between the Neogene (Pliocene) and the Quaternary (Pleistocene) when climate
388 became drier, and many paleotropical and mesic temperate forest taxa disappeared to give
389 way to a similar Mediterranean type vegetation to that which exists today (Rundel et al.,
390 2016). This definitive onset of Mediterranean climate may have intensified habitat filtering and
391 community assembly processes, and adaptive radiations of species from the Cistaceae and
392 Fabaceae families may have occurred (Guzmán et al., 2009). In this case, the HS trait may have
393 been an advantage under Mediterranean-dry conditions, which may be fundamental for
394 determining the current phylogenetic structure of plant communities.

395 Our results support the hypothesis that climate is a determinant evolutionary force in
396 the selection of HS in MB flora, or at least for the Fabaceae family species where the presented
397 results are more evident. HS trait selection could have been the result of the adaptive
398 advantage of being a gap-detecting mechanism that coordinates germination after summer.

399 Daily soil temperature fluctuations that rose in bare soils would break the physical dormancy
400 of seeds to allow subsequent germination (Baeza and Roy, 2008; Santana et al., 2013a). The
401 advantage here would be to predispose seeds to germinate during the wet autumn season,
402 when shorter days and lower temperatures would help to keep soil moisture. It would be then
403 when non-dormant seeds would be imbibed and germinate with enough time for them to
404 establish and grow before the next limiting dry season (i.e., summer). HS also allows the
405 formation of a persistent soil seed bank with numerous viable seeds that overcome plant life
406 to ensure persistence in stressful environments (i.e., seasonal droughts in this case). Indeed
407 this persistent seed bank enables a bet-hedging strategy to be incorporated; i.e., temperature
408 fluctuations only break the dormancy of a proportion of the seed bank, so seeds do not
409 germinate all at once (Venable, 2007). In this way, the risk of seedling establishment failure
410 would spread over time and thus ensure further germinations in following years under more
411 favourable conditions (Fenner and Thompson, 2005). In this sense, it would be interesting to
412 ascertain which of these dormancy dimensions has been more important for HS selection: bet-
413 hedging or synchronous germination. Bet-hedging would be related to high aridity conditions,
414 and synchronous germination to high precipitation seasonality. Our results are restricted to
415 woody perennial species which are expected to live more than 1 year and, thus, synchronous
416 germination is key to overcome the limiting first summer. It is well-known that woody
417 perennial species germinate mainly under long wet conditions, especially Autumn but, for
418 example, germination is nil after summer storms (Santana et al., 2012). A different case would
419 be herbaceous species, which are able to germinate and establish after sudden summer
420 storms. Here the bet-hedging dimension, which avoids losing all seeds in a single unfavourable
421 event, may gain relevance. For this reason, further studies are necessary to ascertain the
422 phylogenetic community patterns of herbaceous species and to compare them with woody
423 species. In temperate ecosystems, for some herbaceous species it has been observed that cold
424 stratification produced by winter conditions is needed to break their physical dormancy, and

425 not only high temperature fluctuations (Van Assche et al., 2002). In relation to herbaceous
426 species, these points remain to be tested in MB ecosystems, but should be considered to
427 interpret our results and future studies.

428 Lack of germination that is exclusively dependent on fire temperature (fire-dependent)
429 in MB species may also support the climate hypothesis as a selective force of HS. For example,
430 the seeds of *Ulex parviflorus* (Fabaceae) and *Cistus albidus* (Cistaceae) significantly germinate
431 under summer conditions in open vegetation gaps (Santana et al., 2013a). The daily soil
432 temperature fluctuations in these gaps in summer reach values of 65-68°C, which suffice to
433 enhance germination in this place and contrast with the temperatures that seeds found under
434 vegetation (Santana et al., 2013a). It is also well-known that a germination flush in these
435 families (Fabaceae and Cistaceae) occurs after disturbances other than fire; e.g., clearings and
436 mechanical brushings (Paynter and Flanagan, 2004; Baeza and Roy, 2008; Santana et al.,
437 2013a). These species are also involved in other different ecological processes from post-fire
438 regeneration; e.g., they are considered good colonisers of open spaces and typically occur in
439 early succession stages in abandoned fields (Tatoni et al., 1994; Gallego et al., 2004; Duguay and
440 Vallejo, 2008). Laboratory studies indicate that simulation of soil temperature fluctuations is
441 effective in breaking the physical dormancy of a significant proportion of dormant seeds from
442 the Fabaceae and Cistaceae families (Baeza and Roy, 2008; Moreira and Pausas, 2012).
443 However, very few works have investigated these thresholds for the wide range of species
444 from these families in the MB.

445 Unlike the few works that have dealt with summer temperature fluctuations, there is a
446 long-standing history of works that have studied germination response to simulated fire
447 temperatures in the laboratory for MB species (See Paula and Pausas, 2008 and references
448 therein). In short, it has been widely observed that maximum germination rates occur at fire
449 temperatures (e.g., >80°C). These observations have led to suggest that fire is a determinant
450 factor in HS trait selection in MB species (Moreira and Pausas, 2012). However, this contrasts

451 with the fact that no relation was herein found between fire activity and phylogenetic
452 clustering on a regional scale. In fact the highest clustering appeared in the most arid areas,
453 characterised by little fire activity and high temperatures. Some explanation of the minor role
454 of fire, compared to climate, in affecting phylogenetic clustering can be found in: (i) dormancy
455 temperature thresholds; (ii) the bet hedging capacity of species. Ooi et al. (2014) classified
456 species with germination after summer temperatures into the functional group of “facultative
457 pyrogenic dormancy release”; i.e., they are able to germinate under both conditions (fire and
458 summer temperatures). These species commit a proportion of *ca.* 20% of seeds to germinate
459 after summer temperature treatments and leave a large seed bank remaining. Conversely, fire-
460 related temperatures (>80°C) produce high germinations rates and a limited proportion of
461 seeds remains dormant. This fact questions the advantage of high post-fire germination rates.
462 Post-fire high seed bank depletion may jeopardise regeneration under extreme drought
463 conditions by killing germinated seeds. If recurrent fires occur, the seed bank may also
464 undergo significant depletions and species extinction. This suggestion falls in the line with the
465 response to recurrent fires observed in MB ecosystems. For example, dense *Ulex parviflorus*
466 gorse forms after a first fire in abandoned crop fields. However, recurrent fires (two and three
467 fires in two decades) do not enhance the presence of this species, but lead to a drastic
468 reduction (Santana et al., 2013b). A similar pattern has been found for *Cistus* species in NE
469 Spain with increased abundance after one or two fires, but a drastic decline occurs after three,
470 four and five recurrent fires in three decades (Vilà-Cabrera et al. 2008).

471 It is noteworthy that the phylogenetic structure of the Cistaceae family did not
472 completely follow the expected pattern; i.e., it showed some phylogenetic clustering in
473 northern humid areas. Investigating the phylogenetic structure using isolated traits, as in this
474 study, may not always be an effective approach. Environmental effects on particular traits
475 might not be independent of other traits due to evolutionary trade-offs (Herrera, 1992;
476 Gerhold et al., 2015). This could be particularly relevant in the Cistaceae family, which gathers

477 a significant number of structural traits to face water deficit; e.g., traits to improve water use
478 efficiency and to uptake resources in leafs and roots (Paula and Pausas, 2006, 2011).
479 Therefore, a multivariate approach to gain an in-depth understanding of the evolutionary
480 trade-offs of Cistaceae could be interesting to disentangle this unexpected pattern, as well as
481 the selection of the HS trait and other traits related to drought tolerance. At the same time,
482 interactions with other environmental variables (e.g., soils) may also affect the phylogenetic
483 structure. Further studies may be needed to assess if local endemisms, with specific traits that
484 allow them to live under restricted soil conditions, may affect this pattern. For example,
485 Cistaceae have a significant number of soil-driven vicariants (Mateo and Crespo, 2001), a point
486 that could explain the selection of soil as an explanatory variable in our model, and the large
487 proportion of the variance explained by this variable (17%) regarding the phylogenetic
488 structure of communities.

489 These results for the MB contrast with other Mediterranean type ecosystems, where
490 fire-dependent regeneration has been observed for different species with HS. In Californian
491 chaparral, for example, species from the *Ceanothus* (Rhamnaceae) genus only germinate if
492 seed banks face temperatures generated during fires (Keeley, 1991). In this case, HS has been
493 considered to emerge much earlier as an adaptation to fire than to Mediterranean dry climate
494 under the fiery conditions of the Neogene period (Pausas et al., 2006). Along the same line, the
495 germination response of some species from the *Acacia* genus (Fabaceae) in Australian
496 ecosystems are restricted to fire temperatures over 90°C (Ooi et al., 2014; Cochrane, 2017). In
497 these cases, adaptation to fire seems clearer as species undergo pyrogenic dormancy release,
498 which only commits a proportion of their seed bank and permits bet-hedging mechanism
499 efficiency to be preserved (Ooi et al., 2014). However, it is worth noting that species with fire-
500 dependent regeneration are accompanied by many other species, which display facultative
501 dormancy release and behave as opportunistic species by germinating and colonising open
502 gaps when fire is absent (Keeley et al., 2006; Ooi et al., 2014). So it would be interesting to

503 follow similar approaches to that herein presented in these Mediterranean type ecosystems as
504 they could shed some light to clarify if the post-fire germination flush that occurs in gap-
505 detecting species is an exaptation rather than an adaptation to fire, as suggested for MB
506 ecosystems. This could also provide insight into the evolution of fire-dependent germination:
507 whether it appeared in older lineages prior to the Mediterranean conditions under fiery
508 conditions (as suggested for Chaparral species), or whether it was selected by starting from an
509 adaptation to arid conditions that evolved towards a fire-adaptive trait, which may have also
510 occurred for some Australian species of the *Acacia* genus.

511

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521 **References**

522 Ackerly, D.D., 2009. Evolution, origin and age of lineages in the Californian and Mediterranean
523 floras. *J. Biogeogr.* 36, 1221-1233.

524 Alimadad, A., Salibian-Barrera, M., 2011. An outlier-robust fit for generalized additive models
525 with applications to disease outbreak detection. *J. Am. Stat. Assoc.* 106, 719-731.

526 Baeza, M. J., Roy, J., 2008. Germination of an obligate seeder (*Ulex parviflorus*) and
527 consequences for wildfire management. For. Ecol. Manag. 256, 685-693.

528 Baskin, J.M., Baskin, C.C., Li, X., 2000. Taxonomy, anatomy and evolution of physical dormancy
529 in seeds. Plant Species Biol. 15, 139-152.

530 Baskin, C.C., Baskin, J.M. 2014. Seeds: ecology, biogeography, and evolution of dormancy and
531 germination. Academic Press, San Diego.

532 Bradshaw, S.D., Dixon, K.W., Hopper, S.D., Lambers, H., Turner, S.R., 2011. Little evidence for
533 fire-adapted plant traits in Mediterranean climate regions. Trends Plant Sci. 16, 69-76.

534 Burnham, K.P., Anderson, D.R., 2002. Model selection and multimodel inference: a practical
535 information-theoretic approach. Springer-Verlag, New York.

536 Cavender-Bares, J., Ackerly, D.D., Baum, D.A., Bazzaz, F.A., 2004. Phylogenetic overdispersion
537 in Floridian oak communities. Am. Nat. 163, 823-843.

538 Chamberlain, S., 2016. brranching: Fetch 'Phylogenies' from Many Sources. R package version
539 0.2.0. <https://CRAN.R-project.org/package=brranching> (accessed 15 november 2017).

540 Cochrane, A., 2017. Are we underestimating the impact of rising summer temperatures on
541 dormancy loss in hard-seeded species? Aust. J. Bot. 65, 248-256.

542 COPUT., 1998. Litología, aprovechamiento de rocas industriales y riesgo de deslizamiento en la
543 Comunidad Valenciana. Cartografía temática N°5. V+CD-ROM.

544 Dantas, V.D.L., Batalha, M.A., França, H., Pausas, J.G., 2015. Resource availability shapes
545 fire-filtered savannas. J. Veg. Sci. 26, 395-403.

546 Dormann, C.F., Elith, J., Bacher, S., Buchmann, C., Carl, G., Carré, G., García-Márquez, J.R.,
547 Gruber, B., Lafourcade, B., Leitao, P.J., et al., 2013. Collinearity: a review of methods to deal
548 with it and a simulation study evaluating their performance. *Ecography* 36, 27-46.

549 Duguy, B., Vallejo, V.R., 2008. Land-use and fire history effects on post-fire vegetation
550 dynamics in eastern Spain. *J. Veg. Sci.* 19: 97-108.

551 Fenner, M., Thompson, K., 2005. *The ecology of seeds*. University Press, Cambridge.

552 Gallego, F.J.B., García, M.M.R., García, N.F., 2004. Vegetation dynamics of Mediterranean
553 shrublands in former cultural landscape at Grazalema Mountains, South Spain. *Plant Ecol.* 172,
554 83-94.

555 Gerhold, P., Cahill, J.F., Winter, M., Bartish, I.V., Prinzing, A., 2015. Phylogenetic patterns are
556 not proxies of community assembly mechanisms (they are far better). *Funct. Ecol.* 29, 600-614

557 Guzmán, B., Vargas, P., 2005. Systematics, character evolution, and biogeography of *Cistus*
558 L.(Cistaceae) based on ITS, trnL-trnF, and matK sequences. *Mol. Phylogenetics Evol.* 37, 644-
559 660.

560 Guzmán, B., Lledó, M.D., Vargas, P., 2009. Adaptive radiation in Mediterranean *Cistus*
561 (Cistaceae). *PLoS One.* 4, e6362.

562 Herrera, C.M., 1992. Historical effects and sorting processes as explanations for contemporary
563 ecological patterns: character syndromes in Mediterranean woody plants. *Am. Nat.* 140, 421-
564 446.

565 Jaganathan, G.K., 2015. Are wildfires an adapted ecological cue breaking physical dormancy in
566 the Mediterranean basin? *Seed Sci. Res.* 25, 120-126.

567 Jiménez-Moreno, G., Fauquette, S., Suc, J.P., 2010. Miocene to Pliocene vegetation
568 reconstruction and climate estimates in the Iberian Peninsula from pollen data. *Rev.*
569 *Palaeobot. Palynol.* 162, 403-415.

570 Katoh, K., Kuma, K., Toh, H., Miyata, T., 2005. MAFFT version 5: improvement in accuracy of
571 multiple sequence alignment. *Nucleic Acids Res.* 33, 511-518.

572 Keeley, J.E., 1991. Seed germination and life history syndromes in the California chaparral. *Bot.*
573 *Rev.* 57, 81-116.

574 Keeley, J.E., Fotheringham, C.J., Baer-Keeley, M., 2006. Demographic patterns of postfire
575 regeneration in Mediterranean-climate shrublands of California. *Ecol Monogr.* 76, 235-255

576 Keeley, J.E., Bond, W.J., Bradstock, R.A., Pausas, J.G., Rundel, P.W., 2012. Fire in Mediterranean
577 ecosystems: ecology, evolution and management. University Press, Cambridge.

578 Keeley, J.E., Pausas, J.G., Rundel, P.W., Bond, W.J., Bradstock, R.A., 2011. Fire as an
579 evolutionary pressure shaping plant traits. *Trends Plant Sci.* 16, 406–411.

580 Kembel, S.W., Cowan, P.D., Helmus, M.R., Cornwell, W.K., Morlon, H., Ackerly, D.D., Blomberg,
581 S.P., Webb, C.O., 2010. Picante: R tools for integrating phylogenies and ecology. *Bioinformatics*
582 26, 1463-1464.

583 Kraft, N.J., Adler, P.B., Godoy, O., James, E.C., Fuller, S., Levine, J.M., 2015. Community
584 assembly, coexistence and the environmental filtering metaphor. *Funct. Ecol.* 29, 592-599.

585 Krawchuk, M.A., Moritz, M.A., 2011. Constraints on global fire activity vary across a resource
586 gradient. *Ecology.* 92, 121-132.

587 Lang, R., 1915. Versuch einer exakten Klassifikation der Böden in klimatischer und geologischer
588 Hinsicht. *Internationale Mitteilungen fuer Bodenkunde* 5, 312-346.

589 Lloret, F., Estevan, H., Vayreda, J., Terradas, J., 2005. Fire regenerative syndromes of forest
590 woody species across fire and climatic gradients. *Oecologia* 146, 461-468.

591 Mateo, G., Crespo, M., 2001. Manual para la determinación de la flora valenciana. Monografías
592 Flora Montiberica, Valencia.

593 Malhado, A.C., Oliveira-Neto, J.A., Stropp, J., Strona, G., Dias, L.C., Pinto, L.B., Ladle, R.J., 2015.
594 Climatological correlates of seed size in Amazonian forest trees. *J. Veg. Sci.* 26, 956-963.

595 Moreira, B., Pausas, J.G., 2012. Tanned or burned: the role of fire in shaping physical seed
596 dormancy. *PLoS One* 7, e51523.

597 Murphy, B.P., Williamson, G.J., Bowman, D.M.J.S., 2011. Fire regimes: moving from a fuzzy
598 concept to geographic entity. *New Phytol.* 192, 316–318.

599 Ojeda, F., Pausas, J.G., Verdú, M., 2010. Soil shapes community structure through
600 fire. *Oecologia* 163, 729-735.

601 Oksanen, J., Blanchet, F.G., Friendly, M., Kindt, R., Legendre, P., McGlenn, D., Minchin, P.R.,
602 O'Hara, R.B., Simpson, G.L., Solymos, P., Stevens, M.H.H., Szoecs, E., Wagner, H., 2017. *vegan*:
603 Community Ecology Package. R package version 2.4-2. [https://CRAN.R-](https://CRAN.R-project.org/package=vegan)
604 [project.org/package=vegan](https://CRAN.R-project.org/package=vegan) (accessed 6 August 2017).

605 Ooi, M.K., Denham, A.J., Santana, V.M., Auld, T.D., 2014. Temperature thresholds of physically
606 dormant seeds and plant functional response to fire: variation among species and relative
607 impact of climate change. *Ecol. Evol.* 4, 656-671.

608 Paradis, E., Claude, J., Strimmer, K., 2004. APE: analyses of phylogenetics and evolution in R
609 language. *Bioinformatics* 20, 289-290.

610 Parker, V.T., Kelly, V.R., 1989. Seed bank in California chaparral and other Mediterranean
611 climate shrubland, in: Leck, M.A., Parker, V.T., Simpson, R.L., (Eds.) Ecology of soil seed banks.
612 Academic Press, San Diego, pp. 231-256.

613 Paula, S., Pausas, J.G., 2006. Leaf traits and resprouting ability in the Mediterranean
614 basin. *Funct. Ecol.* 20, 941-947.

615 Paula, S., Pausas, J.G., 2008. Burning seeds: germinative response to heat treatments in
616 relation to resprouting ability. *J. Ecol.* 96, 543-552.

617 Paula, S., Pausas, J.G., 2011. Root traits explain different foraging strategies between
618 resprouting life histories. *Oecologia* 165, 321-331.

619 Pausas, J.G., Bradstock, R.A., 2007. Fire persistence traits of plants along a productivity and
620 disturbance gradient in Mediterranean shrublands of south-east Australia. *Glob. Ecol.*
621 *Biogeogr.* 16, 330-340.

622 Pausas, J.G., Paula, S., 2012. Fuel shapes the fire–climate relationship: evidence from
623 Mediterranean ecosystems. *Glob. Ecol. Biogeogr.* 21, 1074-1082.

624 Pausas, J.G., Keeley, J.E., Verdú, M., 2006. Inferring differential evolutionary processes of plant
625 persistence traits in Northern Hemisphere Mediterranean fire-prone ecosystems. *J. Ecol.* 94,
626 31-39.

627 Paynter, Q., Flanagan, G.J., 2004. Integrating herbicide and mechanical control treatments with
628 fire and biological control to manage an invasive wetland shrub *Mimosa picra*. *J. Appl. Ecol.* 41,
629 615–629.

630 Pérez-Cueva, A.J., 1994. Atlas climàtic de la Comunitat Valenciana. 1961-1990. Generalitat
631 Valenciana, Valencia

632 Pignatti, S., 1978. Evolutionary trends in Mediterranean flora and vegetation. *Plant Ecol.* 37,
633 175-185.

634 Raunkiaer, C., 1934. *The life form of plants and statistical plant geography.* Clarendon Press,
635 Oxford.

636 Raven, P.H., Axelrod, D.I., 1978. *Origin and relationships of the California flora (Vol. 72).*
637 University of California Press, California.

638 Revell, L.J., 2012. phytools: an R package for phylogenetic comparative biology (and other
639 things). *Methods Ecol. Evol.* 3, 217-223.

640 Rubio de Casas, R., Willis, C.G., Pearse, W.D., Baskin, C.C., Baskin, J.M., Cavender-Bares, J.,
641 2017. Global biogeography of seed dormancy is determined by seasonality and seed size: a
642 case study in the legumes. *New Phytol.* 214: 1527-1536.

643 Rundel, P.W., Arroyo, M.T., Cowling, R.M., Keeley, J.E., Lamont, B.B., Vargas, P., 2016.
644 *Mediterranean Biomes: Evolution of Their Vegetation, Floras, and Climate.* *Annu. Rev. Ecol.*
645 *Evol. Syst.* 47, 383-407.

646 Santana, V.M., Baeza, M.J., Maestre, F.T., 2012. Seedling establishment along post-fire
647 succession in Mediterranean shrublands dominated by obligate seeders. *Acta Oecol* 39, 51-60.

648 Santana, V.M., Baeza, M.J., Blanes, M.C., 2013a. Clarifying the role of fire heat and daily
649 temperature fluctuations as germination cues for Mediterranean Basin obligate seeders. *Ann.*
650 *Bot.* 111, 127-134.

651 Santana, V.M., Baeza, M.J., Marrs, R.H., 2013b. Response of woody and herbaceous fuel to
652 repeated fires in Mediterranean gorse shrublands. *Int. J. Wildland Fire* 22, 508-514.

653 Stamatakis, A., 2006. RAxML-VI-HPC: maximum likelihood-based phylogenetic analyses with
654 thousands of taxa and mixed models. *Bioinformatics* 22, 2688-2690.

655 Suc, J.P., 1984. Origin and evolution of the Mediterranean vegetation and climate in
656 Europe. *Nature* 307, 429-432.

657 Taton, T., Magnin, F., Bonin, G., Vaudour, J., 1994. Secondary successions on abandoned
658 cultivation terraces in calcareous Provence. I- Vegetation and soil. *Acta Oecol.* 15, 431-447.

659 Thompson, J.D., 2005. *Plant evolution in the Mediterranean*. Oxford University Press, Oxford.

660 Trabucco, A., Zomer, R.J., 2009. Global aridity index (global-aridity) and global potential evapo-
661 transpiration (global-PET) geospatial database. CGIAR Consortium for Spatial Information.
662 <http://www.cgiar-csi.org/> (accessed 15 April 2017).

663 Van Assche, J.A., Debucquoy, K.L., Rommens, W.A., 2003. Seasonal cycles in the germination
664 capacity of buried seeds of some Leguminosae (Fabaceae). *New Phytol* 158, 315-323.

665 Van Der Werf, G.R., Randerson, J.T., Giglio, L., Gobron, N., Dolman, A.J., 2008. Climate controls
666 on the variability of fires in the tropics and subtropics. *Global Biogeochem. Cycles* 22.

667 Venable, D.L., 2007. Bet hedging in a guild of desert annuals. *Ecology* 88, 1086-1090.

668 Verdú, M., Pausas, J.G., 2007. Fire drives phylogenetic clustering in Mediterranean Basin
669 woody plant communities. *J. Ecol.* 95, 1316-1323.

670 Vilà-Cabrera, A., Saura-Mas, S., Lloret, F., 2008. Effects of fire frequency on species
671 composition in a Mediterranean shrubland. *Ecoscience* 15, 519-528.

672 Webb, C.O., Ackerly, D.D., McPeck, M.A., Donoghue, M.J., 2002. Phylogenies and community
673 ecology. *Annu. Rev. Ecol. Evol. Syst.* 33, 475-505.

674 Webb, C.O., Donoghue, M.J., 2005. Phylomatic: tree assembly for applied phylogenetics. *Mol.*
675 *Ecol. Resour.* 5, 181-183.

676 Willis, C.G., Baskin, C.C., Baskin, J.M., Auld, J.R., Venable, D.L., Cavender-Bares, J., Donohue, K.,
677 Rubio de Casas, R., The NESCent Germination Working Group., 2014. The evolution of seed
678 dormancy: environmental cues, evolutionary hubs, and diversification of the seed plants. *New*
679 *Phytol.* 203, 300-309.

680 Wood, S.N., 2011. Fast stable restricted maximum likelihood and marginal likelihood
681 estimation of semiparametric generalized linear models. *J. Royal Stat. Soc. Series B* 73, 3-36

682 Yu, G., Smith, D., Zhu, H., Guan, Y., Lam, T.T., 2017. ggtree: an R package for visualization and
683 annotation of phylogenetic trees with their covariates and other associated data. *Methods*
684 *Ecol. Evol.* 8, 28-36.

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692 **TABLES**

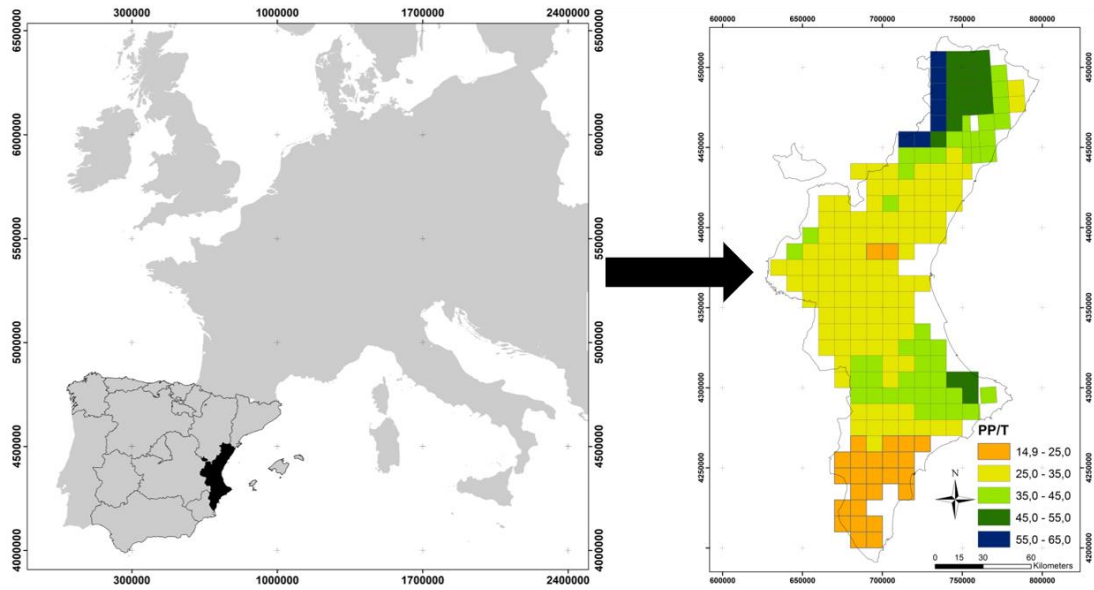
693 **Table 1.** Deviance explained by the GAM models with the included environmental predictors.

694 Model 1 includes fire as a predictor, while Model 2 includes soil as a predictor instead of fire.

	Model 1			Model 2		
	Predictor	Partial Deviance	SE	Predictor	Partial Deviance	SE
(a) HS Trait	All	0.47	-	All	0.46	-
	Space	0.28	0.07	Space	0.27	0.07
	Climate	0.16	0.07	Climate	0.15	0.01
	Fire	0.03	0.03	Soil	0.04	0.07
(b) Fabaceae	All	0.52	-	All	0.50	-
	Space	0.26	0.08	Space	0.24	0.07
	Climate	0.23	0.07	Climate	0.20	0.03
	Fire	0.03	0.003	Soil	0.06	0.07
(c) Cistaceae	All	0.68	-	All	0.69	-
	Space	0.39	0.12	Space	0.32	0.11
	Climate	0.27	0.11	Climate	0.20	0.10
	Fire	0.02	0.012	Soil	0.17	0.08

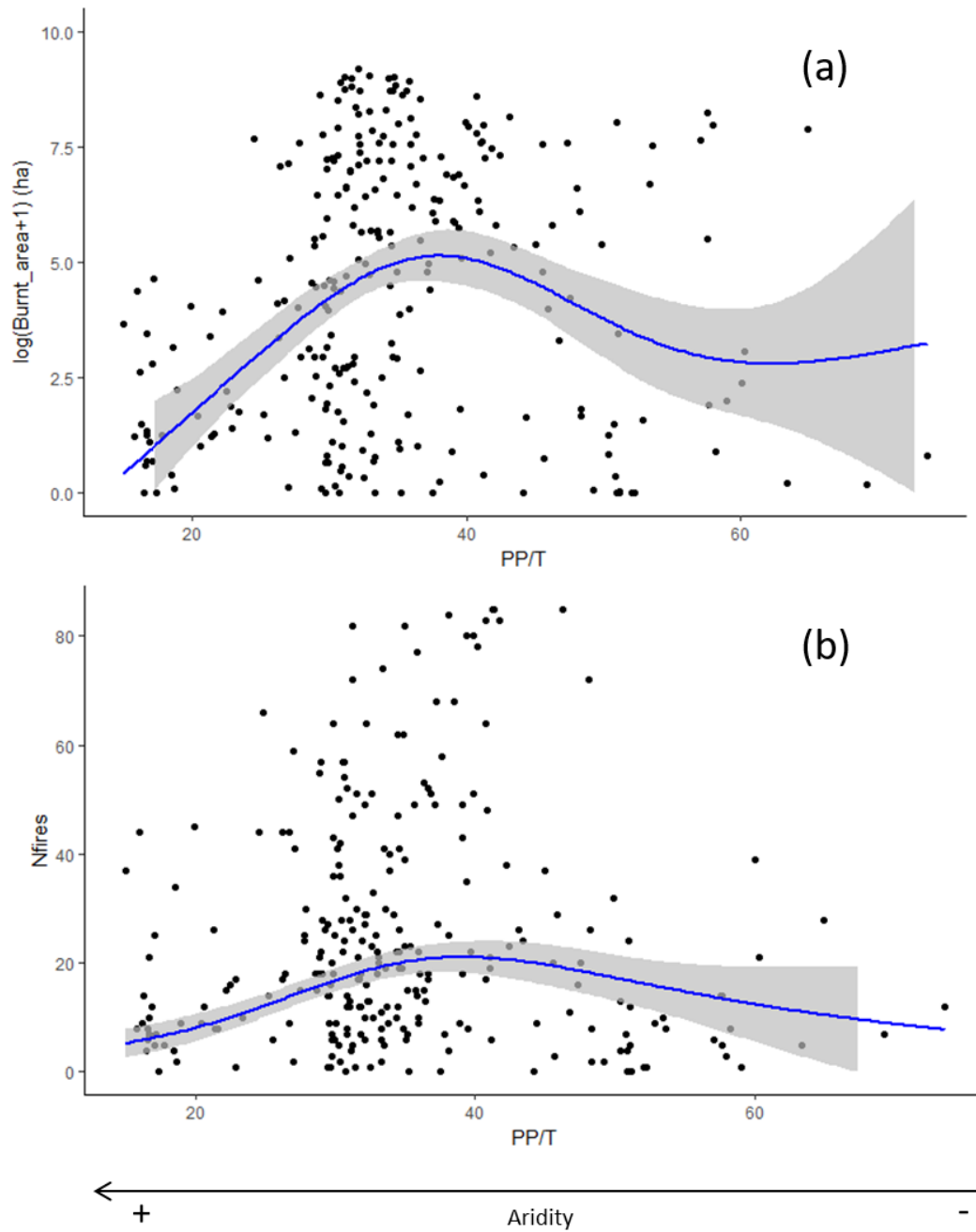
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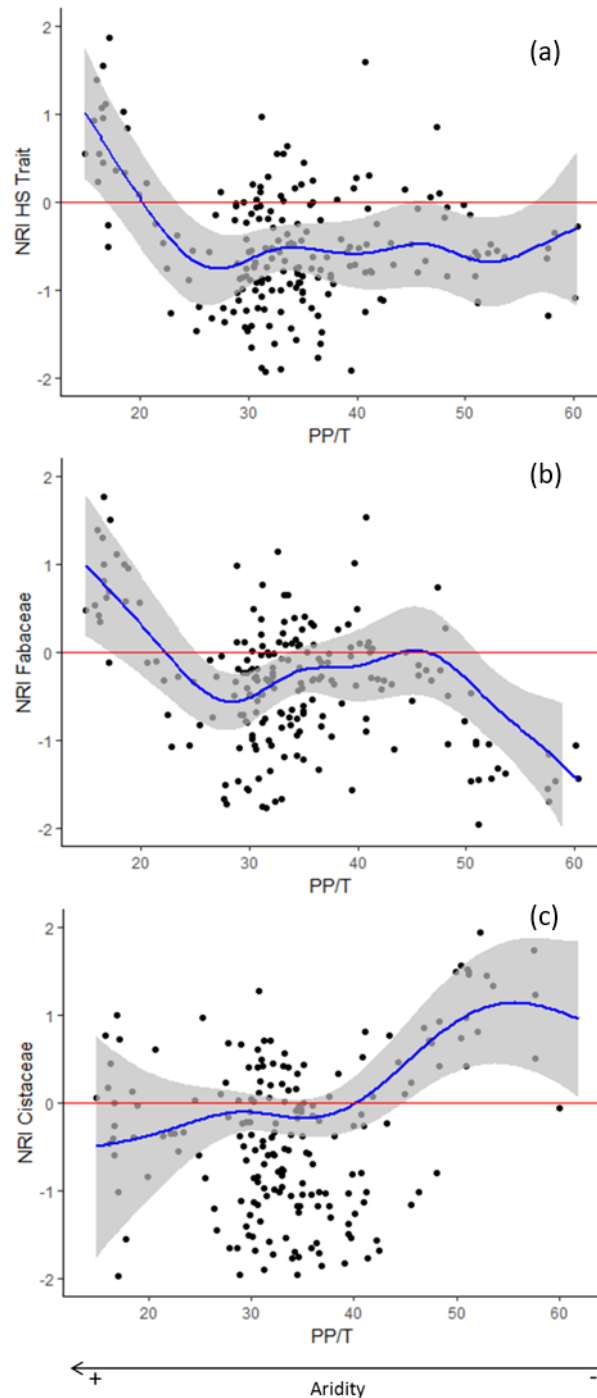
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699 **Figure 1.** Location of the study area (left). The climatic gradient is depicted by the 10×10 UTM
 700 coordinates used in the study (right). The climatic gradient is described by the Rainfall index of
 701 Lang (1915) as quotient PP/T (PP = mean annual precipitation; T = mean annual temperature).
 702 The numbers on the edge of figures indicate UTM coordinates.



703

704 **Figure 2.** Relation of fire activity proxies with the climatic gradient (PP/T) for the Valencia
 705 Region; (a) total area burned; (b) number of fires for each 10×10 UTM coordinate. Blue lines
 706 and CI grey envelopes indicate the GAM models fitted for each proxy using the spatial data
 707 structure to correct spatial-autocorrelation ($P < 0.001$). The climatic gradient is described as
 708 quotient PP/T (PP: mean annual precipitation, T: mean annual temperature).



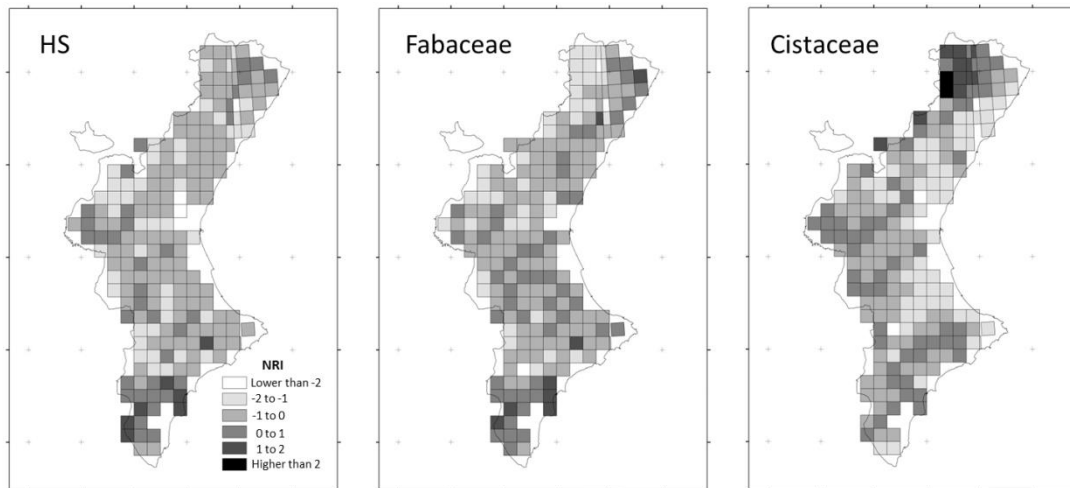
709

710 **Figure 3.** Net relatedness index (NRI) against the climatic gradient depicted in the Valencia
 711 Region (PP/T) for the Hardseededness trait (HS) (a) and the Fabaceae (b) and Cistaceae (c)
 712 families. Blue lines and CI grey envelopes indicate the GAM models fitted for each proxy using
 713 the spatial data structure to correct spatial-autocorrelation ($P < 0.001$). The values above the
 714 red line indicate phylogenetic clustering, whereas those below it denote overdispersion. The

715 climatic gradient is described as quotient PP/T (PP: mean annual precipitation, T: mean annual
716 temperature).

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718



719

720 **Figure 4.** Spatial distribution in the Valencia Region of the Net Relatedness Index (NRI) for the
721 Hardseededness trait (HS) and for the Fabaceae and Cistaceae families. Each square depicts a
722 10 ×10 UTM coordinate.

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