

1 **Strapline:** Thermal niche dimensionality

2

3 **Title:** Thermal niche dimensionality could limit species' responses to temperature
4 changes: insights from dung beetles

5

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20 **Abstract**

21

22 **Aim:** Adequate responses of species to climate changes requires that thermal changes
23 remain compatible across different key biological aspects (e.g. reproduction, feeding
24 and development). However, limits of thermal compatibility to such biological aspects
25 are largely unknown in extant ectotherm groups. To fill this gap, we studied the
26 intraspecific congruence of thermal responses across biological aspects.

27 **Location:** Iberian Peninsula

28 **Taxon:** Scarabaeidae Dung beetles

29 **Methods:** We studied the role of temperature in determining the diel, seasonal, and
30 geographical occurrences of sixteen dung beetle species. We fitted polynomial GLMs of
31 the abundance/occurrence of each species as a function of temperature and alternative
32 predictors for each spatiotemporal scale, using deviance partitioning to explore the
33 relative contribution of temperature. We used the fitted models to estimate realized
34 thermal niche attributes at these three spatiotemporal scales, and assessed their
35 intraspecific congruence through the correlation of niche attributes from different
36 scales.

37 **Results:** We found that temperature has relatively low –but not negligible– explanatory
38 capacity at the three spatiotemporal scales, once alternative predictors are taken into
39 account. More importantly, the estimated thermal responses were largely incongruent
40 across scales, indicating that these species have multidimensional thermal niches.

41 **Main conclusions:** The multidimensionality of thermal niches entails that species’
42 adjustments to fulfil temperature requirements for one biological aspect (such as
43 seasonal ontogenetic cycles) may result in detrimental effects on other elements (*e.g.*,
44 diel activity). These trade-offs could expose individuals to inadequate temperatures at
45 certain moments, reducing populations’ performance. Paradoxically, the weak effects of
46 temperature we found may have severe consequences for species responses to warming
47 if temperature regulates essential aspects of their biology in divergent ways.

48

49 **Keywords:** biological scale, daily activity, geographic distribution, niche
50 dimensionality, phenology, physiological trade-offs

51

52 **Introduction**

53 Temperature is fundamental for the efficient capture and management of the energy that
54 maintains living organisms (Brown et al. 2004). Temperature variations affect the
55 abundance and distribution of species (Angilletta 2009), the variability of ecological
56 systems (Wang et al. 2009), and even the history of life and biodiversity on Earth itself
57 (Schwartzman 1999, Mayhew et al. 2008). Indeed, temperature plays a critical role in
58 controlling key aspects such as species' spatiotemporal distribution, physiological
59 activity, or individual growth rates (Somero 2005, Thackeray et al. 2016, Scranton &
60 Amarasekare 2017, Madrigal-González et al. 2018), among many other things. Here, the
61 effects of temperature on species' geographic distributions and seasonal and diel
62 activities are of particular interest since variation in these aspects can have dramatic
63 consequences for their persistence and ecological performance (Edwards & Richardson
64 2004, Schweiger et al. 2008, Rader et al. 2013). For instances, the ongoing climate
65 changes are drastically modifying the spatial and temporal organization of biodiversity
66 (Chapin III & Diaz 2020), leading to spatial and seasonal decouples of interacting
67 species (Sheldon et al. 2011) and to the disruption of food webs and ecosystem services
68 (Román-Palacios & Wiens 2020). Ecologists and climatologists have accumulated a
69 large amount of evidence on these effects during recent decades, which are especially
70 relevant for ectotherms (Paaajmans et al. 2013). Despite this, how temperature responses
71 integrate across different species' aspects is still largely unknown.

72 Delimiting the actual effect of environmental temperature on the distribution
73 and abundance of species may become difficult when other variables that are either
74 spatially or temporally correlated with temperature are considered simultaneously. For
75 instance, the latitudinal distribution of species in the Northern Hemisphere is associated
76 with historical events and dispersal limitations, whose effects generate geographical
77 patterns that can be confounded with those of temperature variations (Araújo et al.
78 2008, Hortal et al. 2011, Calatayud et al. 2016, 2019). Similarly, the apparent
79 relationships between temperature and either seasonal or diel activities may be indeed
80 conditioned by life-history constraints related to the time required to complete
81 individual development, species' voltinism, the phase in which overwintering occurs,
82 photoperiod limitations, light requirements, and the reliance on solar radiation
83 independently on the environmental temperature (Bradshaw & Holzapfel 2007, 2010,
84 Teder 2020). Hence, assessing the predictive value of temperature in accounting for the
85 spatial and temporal variations in species occurrence and abundance would require

86 considering any alternative variables that could play a significant role in these
87 variations.

88 Experimental setups can help unravel the “true” role of temperature in driving
89 geographical, seasonal, and diel patterns for some model organisms while controlling
90 for other variables (Angilletta 2009). However, experiments based on artificial thermal
91 gradients can subject individuals to new and unrealistic stress conditions, thereby
92 providing overestimated projections of species responses (Guo et al. 2020).
93 Alternatively, one could explore the contribution of temperature using observational
94 data where the variations in temperature and other complementary predictors are
95 decoupled. For example, the effects of temperature and solar radiation can be teased
96 apart using diel activity from consecutive days that showed substantial variations in
97 temperature (*i.e.*, while presenting almost equal insolation). Similarly, temperature and
98 day length effects can be teased apart using seasonal data along steep temperature
99 gradients, with nearly equal day lengths (such as elevational gradients). Finally, the role
100 of temperature in determining the species’ distribution can be assessed by comparing
101 geographical areas with different temperature regimens. That is, if temperature is an
102 important variable, we should find similar responses under different background
103 temperatures.

104 The relevance of temperature in accounting for the spatiotemporal variation in
105 species occurrence and abundance may thus be estimated from observational data,
106 comparing the results from including or not alternative predictors to account for
107 complementary causal factors. Temperature will stand out as a relevant factor across
108 different biological scales if its association with several species' responses is high
109 throughout different spatiotemporal dimensions, but also if such responses are
110 congruent across dimensions for each species. Herein we call intraspecific congruence
111 to the within-species similarity in the thermal responses to diel, seasonal, and
112 geographical gradients. High degrees of such intraspecific congruence would support
113 the universal and homogeneous role of temperature in delimiting species' occurrence
114 and abundance. Note here that expectations are that different mechanisms are behind the
115 response to temperature variations associated with geography, seasonality, and diel
116 rhythms. For instance, daily temperature variations should also be related to changes in
117 light or other environmental factors that can generate behavioural, endocrine, and
118 physiological diel rhythms (Levy et al. 2019). In contrast, responses to seasonal
119 temperatures should be associated with the annual rhythms and the need to synchronize

120 life history phases with seasonal variations in climate (Saunders 2020). On the other
121 hand, responses to geographical variations in temperature should relate to local
122 adaptation processes acting at the population level, and likely involving the above-
123 mentioned individual tolerances and ontogenetic timing, as well as other essential
124 species attributes (Sunday et al. 2019).

125 Despite these differences, a certain level of intraspecific congruence in the
126 responses would indicate the consistent role of temperature as a holistic and predictable
127 driver of key biological aspects. Such within-species congruence would be evident, for
128 example, if species occurring in colder regions are also active during colder periods of
129 the year and at colder hours of the day in areas of a milder climate. Such hypothesized
130 thermal intraspecific congruence is fundamental to respond adequately to global
131 warming, as decoupling responses across different spatiotemporal gradients may expose
132 local populations to critical temperatures, thus compromising their long-term
133 persistence. For instance, if seasonal and diel responses to temperature are decoupled,
134 species might not be able to adjust seasonal cycles as much as it would be necessary to
135 prevent individuals from facing critical temperatures during diel activities. Following
136 this line of evidence, studying the similarity of thermal responses between species
137 across evolutionary lineages is also important because a marked phylogenetic signal in
138 thermal niches would also point to the relevance of temperature changes. That is, if
139 thermal adaptations are evolutionarily conserved, species might present limited ability
140 to modify their thermal responses, being unable to cope with climate warming. Should
141 this be true, phylogenetic biases in the potential effects of climate warming would also
142 be expectable. Despite the relevance of studying the consistency of thermal responses
143 across spatiotemporal gradients and evolutionary lineages, integrative studies on this
144 topic are lacking.

145 Here we study the thermal responses associated with geographical, seasonal,
146 and diel temperature variations using several temperate dung beetle species as a model
147 system. Some dung beetle species are capable to self-regulate their body temperature
148 and produce heat, a faculty that depends to some extent on their body size (Verdú &
149 Lobo 2008, Verdú et al. 2012), a physiological adaptation directly linked to the need for
150 a quick dispersal response to exploit an ephemeral resource. Besides, they feed on dung
151 from domestic and wild animals, participating in nutrient cycling and seed dispersion
152 (Milotić et al. 2019), thus providing important ecosystem functions. These

153 characteristics make dung beetles an ideal and important group to study thermal
154 responses.

155 Specifically, we evaluated the responses of dung beetles to changes in
156 temperature associated with: (i) diel rhythms across three consecutive days with
157 contrasting temperatures; (ii) seasonal rhythms across six sites located at different
158 elevations; and (iii) geographical ranges along five river basins in the Iberian Peninsula
159 (Fig. 1). We hypothesized that if temperature is the main factor determining the activity
160 and distribution of dung beetles, its effect should be observed along the three considered
161 spatiotemporal gradients, and its relevance would be higher if the effects of other
162 alternative and/or complementary factors are low. Furthermore, intraspecific
163 congruence in the different species' thermal responses to diel, seasonal, and
164 geographical changes would be expected if the importance of temperature is
165 independent of the spatiotemporal context. On the contrary, a low explanatory capacity
166 of temperature and a lack of intraspecific congruence in its effects across the three
167 spatiotemporal gradients would support a limited and dissimilar role of temperature
168 depending on the spatiotemporal context. Finally, if species are evolutionarily limited to
169 adapt to new thermal regimens, we expect thermal niches to be phylogenetically
170 conserved.

171

172 **Material and methods**

173 *Data origin*

174 We use data on 16 Iberian dung beetle species of the family Scarabaeidae (ten from
175 Aphodiinae and six from Scarabaeinae subfamilies). These species were selected
176 because they occurred in at least 10% of the samples of the three datasets considered,
177 covering different spatial and temporal extents (see below). All considered species
178 (Table 1) are of small body size, with body weights far smaller than 1.9 g (0.2 g at
179 most), the threshold from which endothermy is thought to appear in this group of
180 beetles (Verdú et al. 2006). All the considered specimens can be unambiguously
181 ascribed taxonomically except in the case of *Onthophagus (Palaeonthophagus) vacca*
182 (Linnaeus, 1767) and *O. (Palaeonthophagus) medius* (Kugelann, 1792), which are
183 cryptic species that can be unambiguously differentiated by using molecular data (Roy
184 et al., 2016). Knowing that the two species co-occur in the survey area (unpublished
185 data) we have decided to consider these specimens as belonging to the *vacca-medius*

186 complex. Temperature–occurrence associations for all these species were examined
187 along: (i) five geographical areas of similar extent but different temperature regimes
188 within the Iberian Peninsula (geographical dataset or GD); (ii) six sites placed across a
189 steep elevational range in Central Iberia, and sampled during the same dates but
190 differing in their environmental temperatures (seasonal dataset or SD); and (iii) three
191 consecutive days with similar daily variations but different weather conditions in a
192 single locality near the centre of the same elevational range (diel dataset or DD).

193

194 *Geographical Dataset.* The GD is divided into five study areas, corresponding to the
195 major river basins of the Iberian Peninsula (Ebro, Duero, Tajo, Guadiana, and
196 Guadalquivir; limits extracted from HydroBASINS data available at
197 www.hydrosheds.org, Lehner & Grill 2013, Fig. 1a). These natural areas were used
198 since they are delimited by marked geomorphological boundaries, which are expected to
199 act as dispersal barriers. Furthermore, they are relatively similar in extent (areas ranging
200 from 5.6×10^4 to 9.7×10^4 km²) and almost follow a latitudinal gradient, hence showing
201 contrasting environmental temperatures (Fig. 1b). In each of these basins, we collected
202 all georeferenced occurrences of the selected species from GBIF (www.gbif.org,
203 accessed May 2020) and additional published sources (Hortal & Lobo 2011). As this
204 kind of data is biased due to historically uneven sampling effort (Lobo et al. 2018), the
205 occurrences were pooled within UTM grid cells of 10 x 10 km spatial resolution. This
206 grain was selected because it corresponds to the effective resolution of most of the
207 occurrence information in the dataset, and it is appropriate to avoid the effects of
208 oversampled localities while retaining reasonable climatic detail. The frequency of each
209 species' occurrence data in temperature bins of 1°C (ranging from -3 to 20°C, n=24) was
210 calculated for each river basin ($24 \times 5 = 120$), and these figures were used as dependent
211 variables in the subsequent regression analyses.

212

213 *Seasonal dataset.* Six sites along an elevational gradient located in the Sierra de
214 Guadarrama (Central Spain) (Fig. 1a, Espinoza 2016) were used to explore the effect of
215 temperature variations in SD. Elevations ranged from 755 to 1900 m a.s.l., separating
216 sites at intervals of approximately 200 m a.s.l.. Each survey site was sampled
217 approximately every three weeks, totalling fourteen times from May 2012 to June 2013.
218 We chose this elevation gradient because these sites showed considerable variations in
219 temperature during the whole period of the surveys (Fig. 1b). The sampling protocol in

220 each periodical sample consisted of five pitfall-traps baited with fresh cattle dung.
221 These were separated around 30 m from each other. Traps were placed in open habitats
222 to avoid potential habitat and shadow effects and were exposed for 48 h. The
223 individuals recorded in these traps were pooled together, obtaining an estimation of
224 each species' abundance per elevation site and date ($6 \times 14 = 84$), which were used as
225 response variables in subsequent statistical analyses.

226

227 *Diel Dataset.* Temperature effects on diel activity were assessed using dung beetle data
228 from a grassland located next to El Ventorrillo MNCN field station, in the Sierra de
229 Guadarrama at an approximate elevation of 1500 m a.s.l. (Fig. 1a). This locality was
230 chosen as it shows a high diversity of dung beetles (between 30 and 40 species
231 belonging to the considered subfamilies; Cuesta & Lobo 2019). We sampled three
232 consecutive days (April 28th–30th 2015) that showed contrasting temperatures, with
233 around 8 °C of difference between the mean temperatures of the coldest and the hottest
234 days (Fig. 1b). Each day, ten pitfall traps baited with fresh cattle dung were distributed
235 around a circumference of approximately 50 m. of radius (*i.e.*, traps were at least 30 m
236 apart from each other). Since we intended to measure flight activity during short
237 periods, the bait was enveloped within a piece of nylon stocking to prevent the beetles
238 to remain within the dung bait across different sampling events. We checked all traps
239 every 30 min. from dawn to dusk (approximately from 7:30 am to 7:00 pm, $n=23$),
240 collecting all individuals to subsequently identify them in the laboratory. Traps were
241 also checked during the night to discard nocturnal activity. Individuals from the ten
242 traps were pooled together, obtaining an estimation of the abundance of active
243 individuals from each species each 30 min ($23 \times 3 = 69$), which were further used as
244 dependent variables.

245

246 ***Temperature measures and alternative correlates***

247 Temperature measures were obtained from different standardized methods for each one
248 of the different spatio-temporal scales considered. For the *Geographical Dataset*, we
249 obtained mean annual temperatures at a 30 sec resolution from the WorldClim database
250 (see www.worldclim.org, Hijmans et al. 2005). We preferred mean annual temperatures
251 over monthly average figures since the precise seasonal activity over the complete study
252 area was unknown for most of the species. Nevertheless, spring and autumn
253 temperatures (the seasons when phenological peaks occur for most species) were

254 positively correlated with mean annual temperatures (Pearson's $r = 0.99$ and 0.97 ,
255 respectively), so we assume that mean annual temperature is a reasonable proxy for both
256 of them.

257 For the *Seasonal Dataset*, we set up a temperature data logger in each of the
258 elevational points during the whole period of the study. This device was placed in the
259 shadow at one meter from the ground to escape from extreme temperatures due to
260 insolation, mimicking the meteorological stations on which WorldClim data are based
261 on. Temperature was recorded each 10 min. and we used the mean daily temperature
262 when pitfall-traps were active.

263 In the *Diel Dataset* case, temperature measurements were taken using five data
264 loggers placed in the study site just in the centre of the circumference formed by the
265 traps. Data loggers were placed to recover temperature measurements from the different
266 microclimatic conditions available for dung beetles: two at one meter over the ground,
267 in the sun and shadow; another two directly on the ground, also both in the sun and
268 shadow; and one buried at 10 cm depth. Preliminary results showed that the mean
269 temperatures from the data logger placed on the ground in the sun were those that best
270 correlated with the species' diel activity, so we used these measurements in subsequent
271 analyses. Temperature was recorded each minute, and average temperatures during the
272 30 min before traps were checked were used as predictors.

273 As previously stated, the effects of temperature measurements might be
274 overestimated due to its collinearity with other factors with which it shares spatial (in
275 the case of GD) or temporal structure (in the case of SD and DD). We quantified this
276 potential overestimation effect by using different “contrast variables”, alternative
277 predictors which are often partly correlated with temperature but are either measures or
278 proxies of other causal factors for dung beetle spatial and temporal responses. These
279 alternative predictors were temperature availability and survey effort in the case of GD,
280 day of the year for SD, and hour of the day in the case of DD. The effect of temperature
281 on the frequency of occurrence (GD) or abundance (SD and DD) that is independent of
282 these contrast variables was assessed as the “pure” effect of temperature variations that
283 is independent of the range of temperatures available (GD), the period of the year (SD),
284 and the hour of the day (DD) (see analytical methods below).

285 Temperature availability for each basin is the relative frequency of 10 x 10 km
286 UTM cells in each 1°C temperature bin. This variable aims to represent the thermal
287 spectrum available in each basin. Hence, a high explanatory capacity of this variable on

288 the frequency of occurrence of a species would imply that the apparent thermal
289 preference of this species can be simply because its spatial pattern of occurrence mimics
290 the distribution of temperatures in the analysed basin. Further, the typical correlation
291 between the observed pattern of occurrence of a species and the spatial distribution of
292 survey effort can also generate spurious correlations between species' frequency and
293 temperature in each basin. This potential error source was considered here by
294 calculating the relative frequency for each 1°C temperature bin of all dung beetle
295 records included in the formerly mentioned databases and pooled within the 10 x 10 km
296 UTM cells. Nevertheless, we found that this estimation of survey effort and temperature
297 availability were highly and positively correlated in all basins (Pearson's r ranging from
298 0.97 to 0.99), since the most frequent temperatures have been also surveyed more often,
299 which implies that the surveys are randomly allocated within the available temperatures.
300 Consequently, we discarded using survey bias as a contrast variable, considering that
301 the effect of temperature availability also includes differences in survey effort. In the
302 case of SD data, the day of the year was obtained by first ordering the available dates
303 from the day corresponding to the summer solstice (June 21th = 0 or 360), to
304 subsequently convert these figures into radians and obtaining two circular variables by
305 calculating their cosine and sine values. Thus, the summer-winter oscillation is
306 represented by the cosine of the date and oscillates from 1 to -1, whereas the spring-
307 autumn transition is represented by the sine of the date scale 1 to -1. Finally, the hour of
308 the day (DD data) is simply codified as the number of minutes from dawn.

309

310 *Statistical analyses*

311 *Explanatory capacity of temperature*

312 We first explored the independent capacity of temperature to explain variations in dung
313 beetle data in GD, SD, and DD. For each dataset, we conducted Generalized Linear
314 Regression Models of the relative frequency or the abundance of each species as a
315 function of temperature values. All data coming from the five basins (in GD), the six
316 elevational sites (SD), and the three days (in DD) were considered at the same time in
317 each one of the three models. A curvilinear quadratic function of temperature was
318 included in all the cases to account for the typical unimodal performance curves of
319 ectotherms (Huey & Kingsolver 1989). A negative binomial error distribution for the
320 dependent variable was assumed to avoid overdispersion issues associated with the
321 Poisson error distribution (Blasco-Moreno et al. 2019), and it was related to the set of

322 predictors via a logarithmic link function. It is important to note that we did not include
323 a term in the models to account for the different spatial (*i.e.*, basins and elevations) and
324 temporal (*i.e.*, days) units. By doing so, we were ignoring other factors that may affect
325 the distribution and activity of dung beetles, besides temperature and the contrast
326 variables analysed. However, this allows us to tease apart the effects of temperature and
327 the formerly mentioned contrast variables with a spatial and temporal structure while
328 avoiding model overparameterization. Hence, we fitted three supplementary GLMs
329 representing different hypotheses regarding the importance of temperature: (i) a full
330 model where both temperature and contrast variables are included altogether, (ii) a
331 model including only these contrast variables, and (iii) a null model where only the
332 intercept was included. We assumed a linear relationship between the density of
333 occurrence and temperature availability (GD); whereas in SD and DD, we assumed
334 curvilinear relationships between abundance and contrast variables by including a
335 quadratic term of both the number of minutes from dawn, and date sine and cosine. We
336 used a deviance partition approach (Legendre 1993, see also Calatayud et al. 2019 for
337 the same approach) to calculate the deviance explained by each set of variables alone
338 (*i.e.*, temperature *vs.* contrast variables; herein, total pseudo R^2) and once accounting for
339 the collinearity with other variables (herein, partial pseudo R^2). Model performance was
340 assessed using the Akaike Information Criterion corrected for small sample size (AICc).

341

342 *Thermal niche attributes*

343 Deriving thermal niches from occurrence data typically provides a partial description of
344 the whole potential response of the species (Sánchez-Fernández et al. 2012, Saupe et al.
345 2018). However, occurrence-based thermal niches may nevertheless be characterized by
346 different attributes such as the optimum temperature and niche breadth (Gouveia et al.
347 2014, Löffler & Pape 2020, Fig. 2). Each species' temperature optimum was assessed by
348 fitting quadratic curves in a GLM and calculating the maxima as their inflection point
349 (see Villén-Pérez & Carrascal 2015 for a similar procedure). Thermal niche breadth was
350 also obtained as the area under the curve of these fitted curves. Fitted values were
351 normalized to reach a maximum value of one to make calculations comparable among
352 datasets and species.

353 We evaluated the intraspecific dissimilarity in the thermal niches across different
354 spatial and temporal scales, herein called “thermal lability”, using data from the
355 different study units used in each dataset; that is, between river basins, elevation sites,

356 and days (Fig. 2). Thermal lability between pairs of units was measured using the
357 Simpson index as follows:

$$358 \quad S = \frac{\min(b,c)}{a + \min(b,c)},$$

359 where a represents the area under the curves where both curves overlap and b and c
360 represent the independent areas under the curves in study units (see Fig. 2). The larger
361 the overlap between the curves obtained at different scales, the smaller the thermal
362 lability will be. We computed this index for all pairs of units in each dataset (*i.e.*, for
363 each pair of basins, each pair of elevations, and each pair of days) and then considered
364 the maximum dissimilarity among all pairs from the same dataset, as this measure will
365 provide a more realistic estimate of the potential thermal lability of each species.

366

367 *Intraspecific congruence in thermal niches*

368 The different nature of temperature measures avoids reliable comparisons of the thermal
369 niches derived from the three datasets (*i.e.*, GD, SD, and DD) for each species
370 individually. For this reason, we evaluated a generalized intraspecific congruence using
371 rank correlations in the niche attributes of all species between each pair of datasets.

372 Hence, the intraspecific niche congruence was assessed using Spearman rank
373 correlations between the deviance explained by temperature (*i.e.*, both for the total and
374 partial pseudo R^2 s), as well as the obtained temperature optima, thermal niche breadths
375 and thermal labilities for each pair of datasets. In addition to these descriptors, we
376 explored the congruence in the overlap of the thermal niches estimated for the same
377 species but from different datasets. To do this, we examined whether interspecific
378 thermal niche dissimilarities were correlated between the different datasets. We
379 computed dissimilarities between the models' normalized fitted values where the
380 temperature was the only explanatory variable using the Simpson index as previously
381 explained, but in this case between pairs of species (see also Fig. 2). By doing so, we
382 created a thermal niche pairwise dissimilarity matrix for each dataset. Then, we
383 conducted Mantel tests based on Spearman's ρ correlation coefficient to assess the
384 relationship between dissimilarity matrices obtained from the different datasets.
385 Significance was evaluated by comparing observed ρ coefficients with 999 null values
386 obtained from different permutations of the dissimilarity matrices.

387

388 *Phylogenetic signal*

389 The potential lability of thermal niches was also assessed from an evolutionary point of
390 view. In this sense, a marked phylogenetic signal would indicate both potential
391 evolutionary constraints for temperature variation responses, and phylogenetically-
392 structured effects of global warming. We reconstructed a Bayesian phylogenetic
393 hypothesis for 18 species present in our datasets based on two mitochondrial (COI and
394 COII) and one nuclear markers (28S RNA, see Appendix S1 for details on phylogenetic
395 reconstruction). DNA markers were sequenced for this study and retrieved from
396 Genbank (Table S1, accessed in June 2016). Pagel's λ test (Pagel 1999) and Blomberg's
397 K statistics (Blomberg et al. 2003) were used to explore the phylogenetic signal in the
398 five variables considered (total and partial deviance explained by temperature,
399 temperature optimum, thermal niche breadth, and thermal lability). Significance for
400 Pagel's λ was assessed with a likelihood ratio test comparing the negative log likelihood
401 obtained from the original tree topology with the negative log likelihood from a
402 topology transformed to remove the signal (*i.e.*, $\lambda = 0$). In the case of Blomberg's K, we
403 tested for significance by randomizing the labels of the phylogenetic tips and comparing
404 observed and random K values. Finally, we also investigated for phylogenetic signal in
405 the thermal niche dissimilarities for each dataset. To do so, Spearman correlations
406 between thermal dissimilarities and phylogenetic distances were used, assessing
407 significance by comparing observed correlations with null values where the labels of the
408 tips of the phylogeny were randomized. In all cases where tip labels were randomized,
409 p-values were calculated as the proportion of null values being equal or higher than
410 observed values.

411 All analyses were conducted in the R environment (R Core Team 2020), using
412 the *AICcmodavg* package (Mazerolle 2019) to calculate AICc values, the function
413 "sintegral" as implemented in the *Bolstad2* package (Curran 2013) to assess areas under
414 the curves, the *vegan* package (Oksanen et al. 2019) for the Mantel tests, and the
415 *phytools* package (Revell 2012) to calculate Pagel's λ and Blomberg's K.

416

417 **Results**

418 There is an evident gradient in the explanatory relevance of temperature towards higher
419 relevance at progressively larger scales (*i.e.*, geographical > seasonal > diel). Model
420 selection revealed that the full model, including temperature and contrast variables, was
421 the most parsimonious for most species in most datasets (Table 1). As exceptions to this

422 general pattern, in the geographical dataset, the model only including temperature was
423 equivalent to the full model (according to AICc) for one species, and it was also the best
424 supported for another species. In the seasonal dataset, the model only including
425 temperature was the best supported for four species, whereas the model only including
426 contrast variables was equivalent to the full model for just one species. Finally, the
427 model including minute from dawn in DD data was equivalent to the full model for only
428 two species and even better for one species (Table 1). In general, considerable total
429 deviance was explained by the models including temperature and contrast variables
430 (mean pseudo- R^2 s = 0.62, 0.63, and 0.77; ranges = 0.51-0.75, 0.38-0.86, and 0.64-0.86,
431 respectively for GD, SD, and DD; see Fig. 3). Partial regressions revealed that the
432 effects of temperature and contrast variables largely overlap. The deviance
433 independently explained by temperature was low (see Fig. 3). Interestingly, the
434 percentage of deviance explained by temperature decreased from the geographical
435 (mean pseudo- R^2 s = 0.33; range 0.13–0.48), to the seasonal (0.19; 0.05–0.36) and diel
436 datasets (0.08; 0.01–0.20) (see Fig. 3).

437 Thermal niche attributes derived from the different datasets showed little
438 intraspecific congruence. Neither the pseudo R^2 explained by temperature alone nor the
439 total pseudo R^2 were positively and significantly correlated between any pair of
440 datasets, and none of the thermal niche attributes were significantly correlated between
441 the three considered datasets (Table 2). Moreover, Mantel tests showed that
442 interspecific niche dissimilarities were not correlated among the three studied
443 spatiotemporal scales (Table 2). Finally, we did not find a phylogenetic signal for any of
444 these variables in any of the datasets, except in the case of niche breadth for the diel
445 dataset (Table 3).

446

447 **Discussion**

448 The spatial and temporal responses of the studied species show large associations with
449 contrast variables besides temperature, although the influence of temperature on dung
450 beetle occurrence may increase towards larger temporal and spatial scales. This
451 contrasts with our preliminary expectations of a high importance of temperature for
452 dung beetle occurrence and activity based on the known basal ectothermic physiology
453 of the considered species. Further, thermal niches were incongruent across scales for the
454 studied species, and also lacked phylogenetic signal. This suggest that thermal
455 adaptations are highly variable both within and between species.

456 The generally low partial effects of temperature found in our study lead to two
457 important conclusions: (i) the abundance, distribution, and activity of dung beetles are
458 controlled by other factors besides temperature, which are at least partially represented
459 by the *ad hoc* contrast variables used here; and (ii) dung beetle species must have
460 biological mechanisms that provide them with the plasticity required to cope with the
461 temperature variations associated to each spatiotemporal context. Thermoregulation and
462 body heat gain are intimately linked to solar radiation in ectotherms (Angilletta 2009).
463 Indeed, empirical evidence suggest that solar radiation is associated with dung beetles'
464 body temperatures (Bartholomew & Heinrich 1978) and temporal variations in their
465 abundance and species richness (Lobo et al. 1998). Hence, it is likely that this factor is a
466 key environmental control of the diel activity of dung beetles. Regarding annual
467 rhythms, photoperiod seems to be a crucial environmental cue regulating insects'
468 seasonality (Nijhout 1994, Bradshaw & Holzapfel 2007). This is likely the case for dung
469 beetles, given the relatively weak effects of temperature on their phenology found in our
470 study (but see Lumaret & Kirk 1987). Also, the different life-history phases of an insect
471 need to be synchronized seasonally, and these require a minimum amount of time to
472 complete. The development of a dung beetle individual requires from 30 to 80 days
473 depending on the species (Christensen & Dobson 1977, Romero-Samper & Martín-
474 Piera 1995, 2007, Arellano et al. 2017), a time that determines key life-history
475 characteristics such as the number of generations per year or the overwintering phase.
476 These developmental constraints are therefore hard to modify without major
477 evolutionary changes (Teder 2020), thereby limiting the effects of environmental
478 temperature on the seasonal abundance and occurrence of dung beetle species. Finally,
479 many factors contribute to shaping the geographical distribution of dung beetle species,
480 including dispersal limitations (Lobo et al. 2006), historical events (Hortal et al. 2011),
481 or the response to other environmental variables such as precipitation, soil, habitat, or
482 trophic preferences (Lumaret & Kirk 1987, Hanski et al. 1991, Hortal et al. 2001, Lobo
483 & Martín-Piera 2002, Davis & Scholtz 2020). It is important to note that we have not
484 quantified the effects of these variables explicitly, so their inclusion could further
485 weaken the pure effect of temperature on our models.

486 Regardless of the effects of alternative factors, it seems accurate that dung
487 beetles have mechanisms to withstand marked temperature variations, especially those
488 associated with diel and seasonal rhythms. Given the nature of our data and analyses,
489 these mechanisms can be operating either at the population level, at the individual level,

490 or both. At the population level, a high genetic diversity linked to large phenotypic
491 variability can produce the apparently labile thermal responses. That is, as individuals
492 are sorted in time and/or space according to their environmental adaptations,
493 population(s) formed by individuals with different thermal preferences would show a
494 certain level of thermal independence. This mechanism seems more plausible to explain
495 results in the geographical datasets, where river basins can act as dispersal barriers,
496 limiting gene flow and enhancing local adaptation to different temperature regimens
497 (Lenormand 2002). However, it seems less likely that this phenotypic variability alone
498 is responsible for the responses to diel and seasonal temperature variations, where a
499 high gene flow is expected between the individuals and populations that are active at
500 different elevations or days. Physiologically plastic responses allowing individuals to be
501 active at different temperatures seem a more plausible mechanism in this case (Crispo
502 2008). In any case, these two potential mechanisms (phenotypic variability and
503 individual plasticity) are in agreement with the observed lack of phylogenetic signal on
504 species responses to temperature across scales, which indeed suggests a lack of thermal
505 niche conservatism (Gilbert & Miles 2019). The relative contribution of population
506 phenotypic variability and individual plasticity remains elusive, calling for further
507 studies directed to unravel the detailed mechanisms behind the diverse responses to
508 temperature found in our study.

509 Be that as it may, the effects of temperature were significant and not negligible,
510 being larger for species distribution than for seasonal activity, and even smaller for diel
511 activity. The increasing importance towards larger scales may be related to the fact that
512 the effects of temperature on the studied biological aspects are nested. That is, the
513 occurrence in a given location would entail that a species holds the adaptations required
514 to maintain a stable population there, which include physiological and/or behavioural
515 adaptations to cope with the seasonal temperature variations that occur in that locality.
516 In the same way, a population with adults active during a given period of the year
517 should present adaptations to handle the daily temperature variations happening during
518 the days when adults are active. Hence, the hierarchically cumulative effects of
519 temperature across these biological scales may explain why temperature becomes more
520 important for geographic distributions than for temporal activities. Ascertaining the
521 plausibility of this idea requires further investigation of intraspecific responses to daily
522 temperature variations across seasons and seasonal temperature responses throughout
523 different populations placed across the species' geographic distribution.

524 Perhaps the most interesting of our results is the lack of intraspecific
525 congruence in the realized thermal niches across the studied species and spatiotemporal
526 contexts. This means that, for instance, species occurring in colder regions do not
527 appear in colder months nor at colder hours of the day in other regions of our dataset.
528 Recent results corroborate the lack of interaction between seasonal and diel activities in
529 the dung beetles inhabiting a Mediterranean locality because most part of the studied
530 species show a similar diel activity pattern along their seasonal active period (Lobo &
531 Cuesta 2021). This somehow counterintuitive result could be related to the uneven
532 relevance of the alternative variables for the different species and spatiotemporal
533 contexts, which facilitates the decoupling of the thermal responses associated with the
534 distribution and activity of dung beetles. Likely, the processes involved in adult
535 movements, life-history cycles, and population maintenance are differently regulated by
536 temperature, despite their nested nature. In other words, our results suggest that species
537 have multidimensional thermal niches, where each critical biological aspect responds to
538 temperature along a different dimension. Therefore, rather than exerting a universal
539 effect, temperature plays multiple roles in a species' biology and metapopulation
540 dynamics. This lack of intraspecific congruence, together with the low independent
541 effects of temperature found in our deviance partition analyses, indicates that estimates
542 of thermal niches will be, in general, inaccurate and context-dependent. This calls for
543 being particularly cautious when using responses measured at different scales as proxies
544 for future responses to climate change. Predicting the effects of climate change using
545 just a single variable might not be able to account for the whole complexity of the
546 spatial and temporal responses of diversity to a climate change scenario. Our results
547 show, in addition, the difficulties in estimating general thermal niches of species,
548 challenging forecasts of species future dynamics under climate warming based on
549 unidimensional thermal niches (Gvoždik 2018).

550 The partial control of temperature on the activity and distribution of dung
551 beetles may be both a blessing and a curse regarding the effects of climate warming. On
552 the one hand, the apparent thermal lability suggests that temperature increases should
553 not strongly modify neither diel and seasonal activities nor the geographic distribution
554 of dung beetles, likely preventing mismatches with interacting species and the
555 subsequent food chain perturbations. This assumption would contradict the results of
556 studies suggesting moderate or even large effects of climate change on dung beetle
557 distributions (Dortel et al. 2013, Menéndez et al. 2013, Holley & Andrew 2019). On the

558 other hand, the diel, seasonal, or geographical adjustments are among the fastest
559 responses to climate warming (Levy et al. 2019, Duchenne et al. 2020). However, our
560 results suggest that the response towards temperature variations is relatively
561 independent at each spatiotemporal scale. This entails that adjustments to temperature
562 requirements may not be coordinated across key biological aspects. Hence, adjustments
563 to fulfil the temperature requirements for one biological aspect may result in detrimental
564 effects on other aspects, thereby reducing individual and population performance as,
565 *e.g.*, seasonal adjustments may expose individuals to inadequate temperatures during
566 diel activity. In the worst-case scenario, species' incapacity to adjust their temperature
567 requirements by modifying diel, seasonal, and geographical patterns at convenience will
568 increase the likelihood of local extinctions when the individuals are exposed to critical
569 temperatures in their daily or yearly periods of activity. Paradoxically, the partially
570 weak effects of temperature we found may have serious consequences for climate
571 warming if temperature regulates important aspects of species' biology in divergent
572 ways.

573 Overall, our results show that temperature may be less important than other
574 factors in determining dung beetle activity and distribution. Further, the intraspecific
575 incongruences between thermal niches estimated from the geographic distribution and
576 seasonal and diel activities show the complex effects of temperature on key species
577 aspects, pointing to a truly multidimensional nature of thermal niches. Together with the
578 partially weak control of temperature on species activity and distribution, these
579 incongruences may inhibit fast responses to climate warming, potentially exposing
580 individuals to critical, or at least inadequate, temperatures and reducing individual and
581 population's fitness.

582

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587 **Table 1.** AICc values for the models of each species in each dataset. In all cases, we
588 conducted a complete model (Full) including temperature, and the corresponding
589 contrast variables, a model only including temperature (Temp), a model only including
590 contrast variables (Cont), and a null model where no predictor variable was included
591 (Null). Contrast variables were minutes from dawn and its quadratic term for the diel
592 data set; date sine and cosine and their quadratic terms for the seasonal dataset; and
593 temperature availability for the geographic data set. The best models in terms of AICc
594 and the equivalent ones ($\Delta AICc < 2$) are highlighted in bold.

595

Subfamily	Species	Diel				Seasonal				Geographic			
		Full	Temp	Cont	Null	Full	Temp	Cont	Null	Full	Temp	Cont	Null
Aphodiinae	<i>Acrossus depressus</i> (Kugelann, 1792)	176.53	245.53	187.59	264.43	120.25	126.26	120.67	149.35	172.98	173.17	209.50	209.95
Aphodiinae	<i>Agrilus constans</i> (Duftschmid, 1805)	140.21	197.94	145.50	214.29	200.92	210.56	206.92	216.62	211.85	218.76	269.52	289.54
Aphodiinae	<i>Aphodius fimetarius</i> (Linnaeus, 1758)	116.91	162.51	121.10	177.61	201.68	199.58	217.13	214.00	380.68	404.55	419.41	455.12
Aphodiinae	<i>Aphodius foetidus</i> (Herbst, 1783)	42.72	56.56	44.51	64.09	138.39	133.18	146.06	144.19	634.53	679.39	753.81	861.04
Aphodiinae	<i>Colobopterus erraticus</i> (Linnaeus, 1758)	128.81	150.10	136.82	164.96	208.49	251.21	217.59	272.82	343.34	366.05	372.06	410.67
Aphodiinae	<i>Esymus pusillus</i> (Herbst, 1789)	175.62	227.70	171.67	241.45	151.68	163.52	217.18	191.76	147.93	153.05	177.47	180.48
Aphodiinae	<i>Melinopterus sphaelatus</i> (Panzer, 1798)	471.34	534.10	493.63	577.91	307.12	322.42	321.94	343.35	289.63	304.26	354.89	390.33
Aphodiinae	<i>Teuchestes fossor</i> (Linnaeus, 1758)	194.71	256.95	208.75	280.58	89.06	98.58	96.18	116.19	258.00	268.42	304.76	318.01
Aphodiinae	<i>Trichonotulus scrofa</i> (Fabricius, 1787)	144.87	185.24	160.35	207.64	204.82	226.12	216.37	261.94	182.51	186.38	221.98	242.86
Aphodiinae	<i>Volinus sticticus</i> (Panzer, 1798)	305.09	342.35	310.59	371.12	116.05	407.35	121.57	122.05	133.88	132.11	167.53	169.92
Scarabaeinae	<i>Euoniticellus fulvus</i> (Goeze, 1777)	39.24	59.86	49.17	67.31	446.74	473.23	454.04	519.71	285.43	306.31	306.07	352.08
Scarabaeinae	<i>Onthophagus fracticornis</i> (Preysler, 1790)	255.26	325.6	266.11	350.04	188.32	184.52	200.87	201.75	274.14	279.86	317.51	329.60
Scarabaeinae	<i>Onthophagus lemur</i> (Fabricius, 1781)	117.11	153.81	124.15	170.30	200.57	158.52	205.75	174.36	231.64	238.74	290.70	313.01
Scarabaeinae	<i>Onthophagus opacicollis</i> Reitter, 1892	71.03	80.84	76.13	90.69	343.31	355.11	350.70	356.49	207.58	215.60	218.29	251.59
Scarabaeinae	<i>Onthophagus similis</i> (Scriba, 1790)	256.91	342.23	260.48	359.42	612.48	617.79	646.30	658.81	312.48	328.99	363.43	400.08
Scarabaeinae	<i>Onthophagus vacca-medius complex</i>	248.03	300.37	248.30	318.63	285.06	296.59	299.48	318.99	315.14	337.44	352.21	409.05

596

597 **Table 2.** Spearman's ρ correlation coefficients and P-values between the considered
 598 thermal niche attributes measured by the three studied datasets are detailed. DD: Diel
 599 dataset. SD: Seasonal dataset. GD: Geographical dataset. * Results based on Mantel
 600 test.
 601

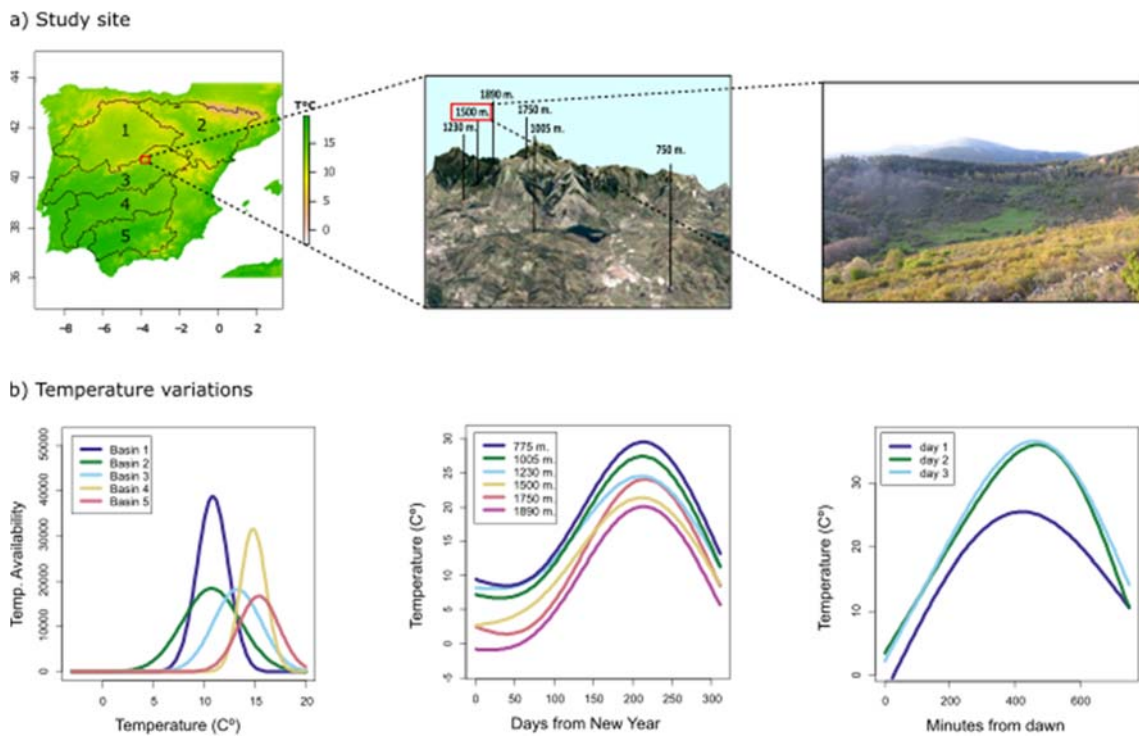
	DD vs SD		DD vs GD		SD vs GD	
	ρ	<i>P</i>	ρ	<i>P</i>	ρ	<i>P</i>
Breadth	-0.074	0.788	-0.385	0.141	0.100	0.713
Optimum	-0.262	0.326	-0.179	0.505	0.394	0.132
Thermal lability	-0.261	0.347	-0.339	0.216	0.132	0.625
Total pseudo R²	0.029	0.914	-0.016	0.953	-0.200	0.456
Partial pseudo R²	-0.561	0.024	-0.440	0.088	0.053	0.848
Niche dissimilarity *	0.260	0.056	0.242	0.051	0.120	0.153

602

603 **Table 3.** Phylogenetic signal in thermal niches attributes for the three studied datasets
 604 (*i.e.*, geographical, seasonal and diel). Significant variables are highlighted in bold. *
 605 Results based on Mantel test.
 606

Variable	Dataset	K	P	λ	P	ρ	P
Breadth	Geographical	0.480	0.150	0	1	-	-
	Seasonal	0.285	0.798	0	1	-	-
	Diel	0.998	0.001	1.096	0.002	-	-
Optimum	Geographical	0.292	0.811	0.002	0.990	-	-
	Seasonal	0.477	0.188	0.361	0.160	-	-
	Diel	0.292	0.710	0	1	-	-
Thermal lability	Geographical	0.475	0.171	0	1	-	-
	Seasonal	0.295	0.900	0	1	-	-
	Diel	0.467	0.228	0.007	0.970	-	-
Total pseudo R²	Geographical	0.287	0.857	0	1	-	-
	Seasonal	0.324	0.710	0	1	-	-
	Diel	0.313	0.726	0	1	-	-
Partial pseudo R²	Geographical	0.367	0.476	0	1	-	-
	Seasonal	0.388	0.398	0	1	-	-
	Diel	0.425	0.350	0	1	-	-
Niche dissimilarity *	Geographical	-	-	-	-	-0.100	0.089
	Seasonal	-	-	-	-	0.200	0.980
	Diel	-	-	-	-	-0.080	0.139

618 **Figure 1.** a) The areas of study in Iberian Peninsula for the geographical, seasonal and
 619 diel dung beetle occurrence datasets (from left to right). Red squares show the position
 620 of the following down-scaled study site. b) Temperature variations in study sites. Lines
 621 correspond with predictions of general additive models (GAM) of: (i) temperature
 622 availability (measured as the number of 10 km² grid cells whose temperature fell within
 623 predefined temperature bins) as function of temperature for the geographic dataset
 624 (left); (ii) temperature as a function of days from New Year and minutes from dawn for
 625 the seasonal and diel datasets respectively. Analyses were computed independently for
 626 each basin, for each elevational site and for each day. GAMs explained an average of
 627 0.90 of deviance across all analyses (median = 0.92, ranging from 0.79 to 0.97).
 628

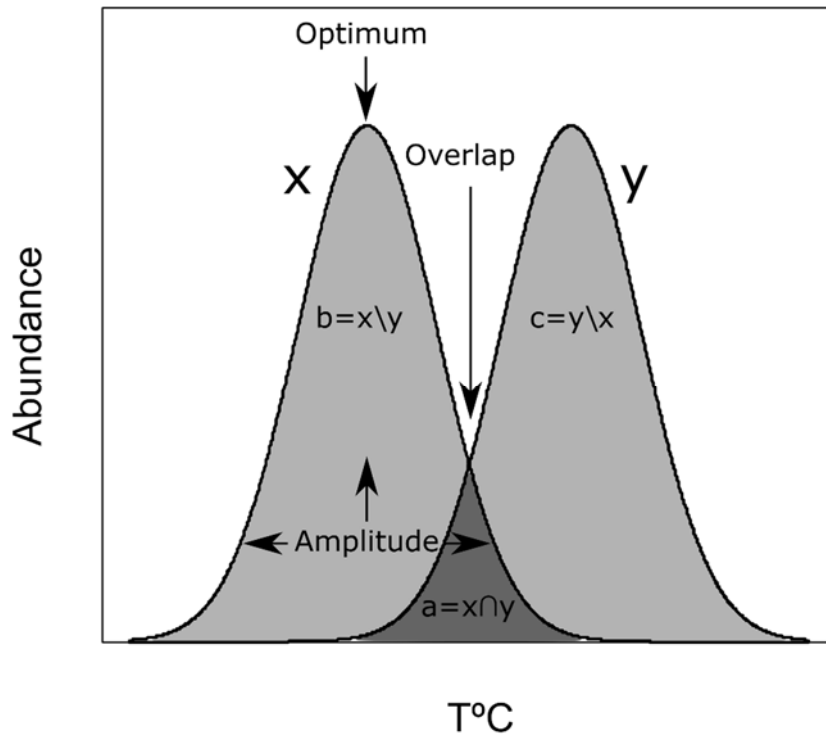


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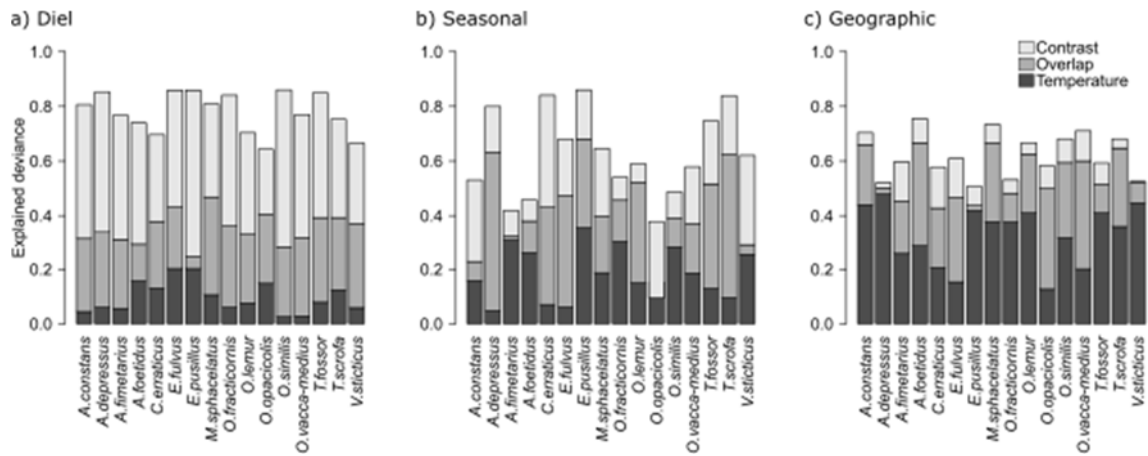
631 **Figure 2. Thermal niche attributes and overlap measure.** x and y represent thermal
632 response curves of two species or a single species in two different study units (*i.e.*, days,
633 elevation, or river basins). From this curve, we obtained the optimum temperature and
634 the niche amplitude. Further, we used the overlap between them (a) and the two
635 independent areas (b and c) to calculate the Simpson's dissimilarity index, as a measure
636 of the congruence between the responses to temperature of the same species at different
637 scales, and of different species within the same scale.

638



639 **Figure 3. Partial regression results.** The deviance explained by temperature alone, the
 640 contrast variables alone, and the overlap between them is shown. The contrast variables
 641 were minutes from dawn and its quadratic term for the diel data set (a); date sine and
 642 cosine and their quadratic terms for the seasonal data set (b), and temperature
 643 availability for the geographic data set (c).

644
 645



646
 647

648 **Data availability:** Data supporting this article is freely available at *figshare* DOI:
649 10.6084/m9.figshare.16451172

650

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- 860
- 861 **Biosketch:** Joaquín Calatayud's research focuses on understanding ecological,
862 evolutionary and historical mechanisms driving species co-occurrence and
863 coexistence.
- 864 **Editor:** Simone Fattorini

Supporting information

Appendix S1

Genomic DNA was extracted from each individual using the BIOSPRINT 15 DNA Kit (Qiagen), following standard manufacturer's protocols for blood, and resuspended in 100 μ l of buffer AE. We used COI Sca F, COI Sca R, COII am Sca and COII B 605 Sea (Villalba et al. 2002) and the universal 28S a y 28S 5b primers to amplify fragments of the mitochondrial cytochrome oxidase I (COI), the cytochrome oxidase II (COII) and the 28S genes. Amplifications for all gene fragments were performed in a 50 μ l reaction containing 39.7 μ l of H₂O, 5 μ l of 10x PCR buffer, 1 μ l of dNTP mix (10 mM), 0.5 μ l of each primer (10 μ M), 0.3 μ l of AmpliTaq® DNA polymerase (Applied Biosystems) and 3 μ l of DNA template. Thermocycling conditions consisted of an initial denaturing step at 94 °C for 4 min, followed by two cycles: (i) a precycle of 5 amplification cycles of 94 °C for 45 sec, 40 °C for 1 min and 72 °C for 1 min, and (ii) a cycle of 35 amplification cycles of 94 °C for 45 sec, 44 °C for 1 min and 72 °C for 1 min, followed by a final elongation step at 72 °C for 10 min and a rapid thermal ramp down to 4 °C. For all reactions, the presence of amplicons of the expected sizes was checked by electrophoresis on a 0.8 % agarose gel. PCR products were purified with the ethanol-precipitation method (Sambrook et al., 1989). Sequencing was performed by Secugen S.L. (Madrid, Spain), using BigDye® and the automated ABI PRISM 3730xl DNA Analyzer. Sequence chromatograms were read and contigs assembled using Sequencher version 4.7 (Gene Codes Corporation, Ann Arbor, MI). All new sequences were deposited in GenBank (see accession numbers in Table S1).

Sequences were aligned in CLUSTALW and MUSCLE, followed by visual inspection using BioEdit (Hall, 1999). Prior to phylogenetic analysis, jModeltest 2.1.1

(Darriba et al., 2012) was used to choose the best-fit model of nucleotide substitution for each of the four genes, and for combined matrices under the corrected Akaike information criterion (AICc). For the COI and COII, HKY was obtained, while Jukes Cantor for 28S. Phylogenetic analyses were performed in a Bayesian framework using BEAST v 2.4 (Drummond and Rambaut, 2007). We established 3 calibrations points based on Ahrens et. al (2014), setting uniform priors with lower and upper boundaries. The calibrations represent the basal split of the following taxa: Aphodiinae (58.7 – 55.8 Million years ago), *Aphodius* (37.2 – 33.9 Mya) and Scarabaeinae (92 – 83.5 Mya). For the age of the rest of the nodes, we set a LogNormal relaxed molecular clock for each gene and let the software estimate the rate from the priors. The MCMC chain ran for 100.000.000 steps, sampled every 10.000 steps. Posterior distribution of all the parameters were checked using Tracer, as well as all ESS values being above 200. We built the tree using Tree Annotator, using the Maximum Clade Credibility implemented method after discarding the first 25% samples as a burn-in.

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Species	28	COI	COII
<i>Aphodius constans</i>	-	AY039372	AY039372
<i>Aphodius depressus</i>	MW342587	MW337285	MW412421
<i>Aphodius erraticus</i>	MW342588	MW337286	MW412422
<i>Aphodius fimetarius</i>	MW342589	MW337287	MW412423
<i>Aphodius foetidus</i>	-	MW337288	MW412424
<i>Aphodius fossor</i>	MW342590	MW337289	MW412425
<i>Aphodius pusillus</i>	MW342591	MW337290	MW412426
<i>Aphodius scrofa</i>	MW342592	MW337291	MW412427
<i>Aphodius sphaelatus</i>	MW342593	MW337292	MW412432
<i>Aphodius sticticus</i>	-	MW337293	MW412428
<i>Euoniticellus fulvus</i>	MW342594	MW337294	MW412429
<i>Geotrupes stercorarius (OUT)</i>	KP419463	AY039377	AY039377
<i>Onthophagus fracticornis</i>	MW342595	-	-
<i>Onthophagus grossepunctatus</i>	MW342596	AY039347	AY039347
<i>Onthophagus lemur</i>	MW342597	AY039353	AY039353
<i>Onthophagus opacicollis</i>	-	MW337295	MW412430
<i>Onthophagus ovatus</i>	MW342598	AY039351	AY039351
<i>Onthophagus similis</i>	MW342599	MW337296	MW412431
<i>Onthophagus vacca</i>	MW342600	AY039359	AY039359
<i>Taurocerastes patagonicus (OUT)</i>	KP419662	GU984611	GU984611

Table S1. GenBank accession numbers of the used sequences. Outgroup species are indicated. In black sequences obtained in this study. In grey sequences retrieved from GenBank.

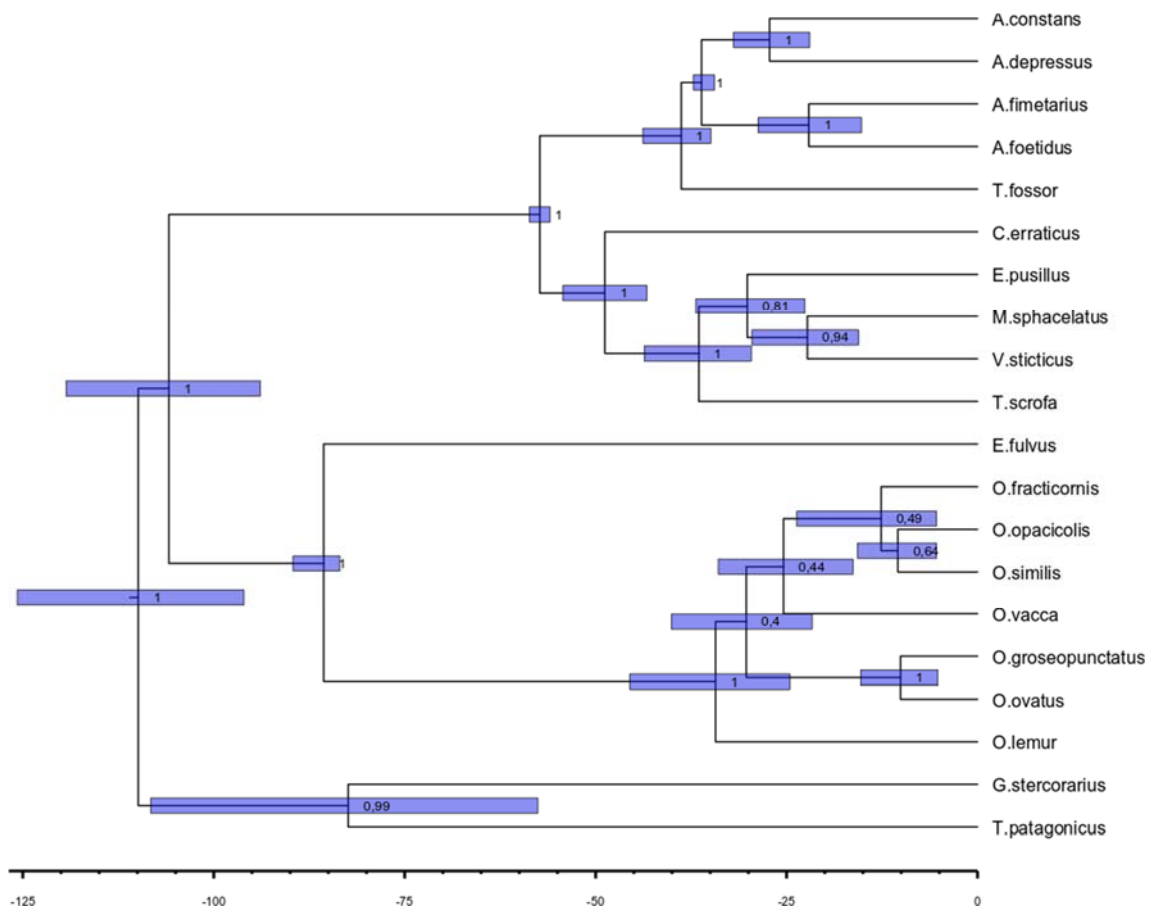


Fig. S1. Bayesian phylogenetic hypothesis for the studied species. Posterior probabilities are provided. Blue bars represent the 95% credible interval around node ages.