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Habitat differences filter functional diversity of low dispersive microscopic animals (Acari: Halacaridae) --Manuscript Draft--

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Abstract:	<p>We are starting to appreciate that microscopic animals are not as widespread as previously thought, but we still ignore to what extent and through which mechanisms the environment selects for specific communities or traits in microscopic animals. We here analyse the functional diversity of marine mite communities living in a seagrass meadow across two habitats: the leaves and the mat. The strictly benthic lifestyle and the conserved morphology of mites allow for unambiguous characterization of their functional traits, while the discrete nature of the two habitats alleviates the uncertainty in their ecological characterization. Our results show that habitat filters the distribution of certain traits favouring a higher diversity, dispersion, and evenness of functional traits in the mat than in the leaves. We further observed temporal variations in the</p>	

	<p>functional diversity of communities, following the changes in biomass and structure of seagrass leaves. However, despite the stark differences between the two habitats, the filtering effect is partial and affects mostly relative species abundances. Our study emphasises the need of moving from a taxonomical toward a functional view of ecological studies of microscopic organisms. This integrative approach is key to achieve a mechanistic understanding of their habitat and distribution patterns.</p>
<p>Response to Reviewers:</p>	<p>We have now replace Garcia-Gomez et al. submitted by Sánchez-Jerez et al. 1999.</p> <p>Sánchez-Jerez, P., Cebrián, C. B., & Esplá, A. A. R. (1999). Comparison of the epifauna spatial distribution in <i>Posidonia oceanica</i>, <i>Cymodocea nodosa</i> and unvegetated bottoms: importance of meadow edges. <i>Acta Oecologica</i>, 20(4), 391-405</p>

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1 **Habitat differences filter functional diversity of low dispersive microscopic animals (Acari,**
2 **Halacaridae)**

3
4 **Running head:** environment affects distribution of mites

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28

29 **Abstract**

30 We are starting to appreciate that microscopic animals are not as widespread as previously thought,
31 but we still ignore to what extent and through which mechanisms the environment selects for
32 specific communities or traits in microscopic animals. We here analyse the functional diversity of
33 marine mite communities living in a seagrass meadow across two habitats: the leaves and the
34 matte. The strictly benthic lifestyle and the conserved morphology of mites allow for unambiguous
35 characterization of their functional traits, while the discrete nature of the two habitats alleviates
36 the uncertainty in their ecological characterization. Our results show that habitat filters the
37 distribution of certain traits favouring a higher diversity, dispersion, and evenness of functional
38 traits in the matte than in the leaves. We further observed temporal variations in the functional
39 diversity of communities, following the changes in biomass and structure of seagrass leaves.
40 However, despite the stark differences between the two habitats, the filtering effect is partial and
41 affects mostly relative species abundances. Our study emphasises the need of moving from a
42 taxonomical toward a functional view of ecological studies of microscopic organisms. This
43 integrative approach is key to achieve a mechanistic understanding of their habitat and distribution
44 patterns.

45

46 **KEYWORDS:** Functional originality; meiofauna; Halacaridae; Posidonia oceanica; n-dimensional
47 hypervolumes; trait ecology

48

49

50 **Introduction**

51 It is unlikely to see a whale gliding in the sky (Adams, 1984) or a bovid grazing on the surface of
52 the ocean (Kavcic et al. 2020). However, as the body size of animals decreases, the probability
53 increases of encountering them in places where *they are not supposed to be*. This is because the
54 realised niche of microscopic animals—namely, where they are actually found—can extend well
55 beyond the set of abiotic conditions that allow positive population growth rates (fundamental
56 niche). These broad ecological ranges are more frequent amongst microscopic animals possessing
57 traits that facilitate long distance dispersal such as dormancy, long-term viability, and
58 parthenogenesis (Fontaneto & Hortal, 2013, Fontaneto, 2019). Similar traits are found, for
59 example, in many species of nematodes (Fonseca & Netto, 2015), rotifers (Fontaneto *et al.* 2008),
60 and tardigrades (Bartels *et al.*, 2020; Kaczmarek et al., 2015). In comparison, some lineages of
61 microscopic organisms are specialised to thrive within narrow ranges of environmental conditions
62 like caves (Mammola et al., 2020a), mountain summits (Hoschitz & Kaufmann, 2004),
63 hydrothermal vents (Zeppilli et al., 2018), and deep terrestrial subsurface habitats (Borgonie et al.,
64 2011). Many of these animals evolved distinct and often convergent traits for these specific
65 conditions. Quintessential examples are microscopic annelids and copepods specialised to feed in
66 the chemocline of certain aquatic caves (Martínez et al., 2019; Worsaae et al., 2019); or mouthless
67 species of nematodes and flatworms living in strict association to prokaryotic symbionts in anoxic
68 marine sediments (Ott et al., 1982).

69 The corollary of these examples is that not only the microscopic body size, but also the
70 presence of certain traits and the interaction between them and the environment, determines the
71 ecological range of microscopic organisms. This is nothing new, as this idea was already grasped
72 in the original formulation of the “*everything small is everywhere*” paradigm, which included the
73 postil “*...but the environment selects*” (Baas-Becking, 1934; Bass & Boenigk, 2011). So we now
74 stand to a point where we know that even broadly distributed and apparently generalist species
75 may not be actually so widespread and tolerant when their habitat preferences are taken into
76 account (or, in other words, that the density of individuals across the distribution range of a given
77 species is not homogeneous as it varies across habitats). But, unfortunately, this filtering effect has
78 proven difficult to quantify, partly due to the lack of data on the relevant traits of many microscopic
79 animals (Giere, 2009) and partly due to the intrinsic problem of measuring relevant environmental
80 variables at appropriate resolutions (Levin, 1992; Potter, Arthur Woods, & Pincebourde, 2013),

81 therefore making it unclear to identify the differences between realized and fundamental niches
82 (Soberón & Nakamura, 2009). These issues have challenged all community-level studies that have
83 so far attempted to directly link functional traits of microscopic animals and their distribution
84 patterns at the relevant scale (Fontaneto 2011). In other words, we know that the environment
85 affects taxonomic and functional diversity in microscopic animals but we ignore to what extent
86 and through which mechanisms the environment selects for specific communities and their traits
87 (e.g. More et al., 2015, Pusceddu et al., 2016; Minor et al., 2017, Semprucci et al., 2018).

88 We here set to examine the effect of habitat on the distribution of microscopic animals by
89 comparing the multidimensional functional space (Blonder et al., 2014; 2018) of assemblages of
90 mites dwelling on a meadow of seagrass [*Posidonia oceanica* (L.)]—a marine plant with a well-
91 studied architecture and growth pattern (Molenaar et al., 2000)—in the Mediterranean. Due to their
92 strictly benthic life mode and easy-to-measure external traits with a clear functional meaning,
93 marine mites are an excellent model system for a similar analysis (Pfungstl et al. 2020).
94 Furthermore, the patchy distribution of seagrass within meadows provides independent replicates
95 of discrete habitats, the leaves versus the matte (i.e., the grid formed by rhizomes, roots, and
96 trapped particles). Because these two habitats present different hydrodynamic regimes (Matteo et
97 al., 1997; Folkard, 2005) and availability of food (Mabrouk et al., 2011; Boudouresque et al.,
98 2016), we expect that they will filter different mites from the pool of species resident in the
99 meadow. We expect that this filter will be evidenced in the community traits, favouring the
100 dominance of more specialised phytophagous or epiphytes feeder species in the leaves, and
101 limiting the presence of generalist detritivores species to the matte. We therefore hypothesise that
102 i) at the community level, there should be higher diversity, dispersion, and evenness of functional
103 traits in the matte than in the leaves. As a corollary of the previous hypothesis, we also expect that
104 ii) at the species level, the higher diversity of traits in the matte will be reflected by the presence
105 of more functionally original species. Furthermore, the annual phenological changes due to the
106 seasonal renovation and decay of seagrass leaves affect nutrient availability (Drew, 1978; Zupo et
107 al., 1997). So, we also hypothesize that iii) differences in functional diversity of mite communities
108 could be related to phenological variation in biomass and structure of *P. oceanica*, particularly on
109 the leaves.

110

111 **Material and methods**

112 Model organism

113 The model organisms selected for this study are marine mites of the family Halacaridae
114 (subsequently referred to as marine mites), a lineage of microscopic arachnids that colonized the
115 ocean from a terrestrial ancestor around 270 million years ago, radiating in different types of
116 marine habitats (Pepato et al., 2018). Due to this terrestrial origin, the body plan of the group is
117 constrained, all forms being restricted to benthic habitats. The impossibility of marine mites to
118 swim or move by any other means than crawling in direct contact with the substrate ensures that
119 the species found in each sample belong to the local community. This feature places marine mites
120 among those animals with a realised niche that is smaller than the fundamental niche, even if they
121 are microscopic: not all available habitats in an area are actually colonised, and the animals are not
122 found in habitats that cannot sustain viable populations. Furthermore, the presence of a hard,
123 hydrophobic cuticle allows for a precise measurement of morphological traits even in fixed
124 material, reducing measure errors. Finally, the conserved morphology of marine mites ensures
125 unequivocal homology assessment of the functional traits. These three properties—movement
126 exclusively by crawling, hard cuticle, and conserved morphology—make marine mites ideal
127 candidates for quantifying the effect of habitat filtering on the distribution and functional diversity
128 of microscopic animals (e.g. Mori et al., 2015, Minor et al., 2017, Pfingstl et al., 2020).

129 Importantly, marine mites are typical and abundant inhabitants of *Posidonia oceanica*
130 meadows (Mari & Morselli, 1990; Durucan, 2018; Durucan & Boyacı, 2018), thriving especially
131 in the vegetated patches (Sánchez-Jerez et al., 1999). This makes it easier to obtain enough
132 specimens for ecological analyses.

133

134 Sampling design

135 As a study area, we selected the exposed seagrass meadow of Cala del Cuartel, in Santa Pola,
136 south-eastern Spain (38° 12' 34.04" N, 0° 30' 19.12" W, WGS84 reference system), consisting of
137 numerous patches at 4–7 m depth separated by bare sandy tongues. Marine mites dwelling in *P.*
138 *oceanica* meadows thrive in seagrass patches and are rarely found in the adjacent bare sand
139 (Sánchez-Jerez et al., 1999). So, in relation to the size and dispersal capabilities of the marine
140 mites, each patch represents a discrete and independent replica of the same habitat within a larger

141 area. The fact that all the patches are within the same bay limits the confounding effect of depth,
142 temperature, salinity, or different exposition to currents.

143 Each patch consists of two compartments representing the two different habitats, the leaves
144 and the matte (Figure 1A). The leaves are exposed to turbulence (Folkard, 2005) and predators
145 (Hovel *et al.*, 2002; Hovel & Fonseca, 2005), as well as affected by changes in length and thus of
146 abundance of epiphytic algae and epifauna (Mabrouk *et al.*, 2011), which potentially represents
147 the main source of food for the mites (Pugh & King, 1985a). In contrast, the matte is a sheltered
148 habitat offering a high and constant availability of detritus throughout the year (Mateo *et al.*, 1997).

149 We performed four sampling campaigns between December 2015 and August 2016. In
150 each campaign scuba divers sampled these two habitats (leaves and matte) in six randomly selected
151 patches of 400 cm² of *Posidonia oceanica* (4 sampling campaign x 6 patches x 2 habitats, totalling
152 48 samples). In each patch, leaves were collected first by cutting them at the ligulae level, while
153 the surface of the underlying matte was collected by scraping the upper 2 cm layer into a separate
154 container.

155 Meiofauna from each sample was extracted combining the magnesium chloride and the
156 ‘bubble and blot’ decantation techniques to ensure the recovery of all species of marine mites
157 (Higgins & Thiel, 1988; Sørensen & Pardos, 2008). The selected mesh size was 62 µm to collect
158 both juveniles and adult forms. Each sample was bulk fixed using 7% formaldehyde in the field.
159 All studied material has been deposited at the Laboratory of Meiofauna at the Universidad
160 Complutense de Madrid.

161 For each leaves sample, as a proxy for food availability, we measured the average length
162 of the leaves, calculated as the distance from the ligula to the apical end of all the complete leaves.
163 Length of the leaves is known to correlate with the abundance of epiphytic organisms (Malbrouk
164 *et al.*, 2011). For each matte sample, as a proxy for food availability, we directly measured the
165 percentage of organic carbon using the approach by Walkley & Black (1934). Furthermore, we
166 inferred habitat availability as the dry weight of leaves or matte divided by the total volume of the
167 habitat, which varied in the leaves (Average leaf length * 20 cm x 20 cm) and was constant in
168 the matte (2 cm x 20 cm x 20 cm).

169

170 Species identification and morphological traits measurement

171 Mites were sorted using a MOTIC[®] SMZ-168 stereoscope, whole-mounted in a modified Hoyer's
172 medium (Mitchell & Cook, 1952), and assigned to species and developmental stages by inspecting
173 relevant morphological characters with a light microscope equipped with Nomarski optics and an
174 Olympus DP70 camera. We used the keys by André (1946) and Green and MacQuitty (1987), as
175 well as the available literature (Bartsch, 1991, 2000, 2001; Morselli, 1980).

176 For each species, we examined 13 morphological traits related to body size and shape, the
177 ability to withstand the water currents, and trophic specialisation (Table 1). Body size and shape
178 measures were taken on all 502 well-preserved specimens from our samples (Table 2). The traits
179 were estimated separately from adults and juveniles (larval or nymphal stages), as different life
180 stages exhibit different ecological preferences and dispersal capabilities even within the same
181 species (Bartsch, 2002; Somerfield & Jeal, 1995; 1996). The other traits, species-specific and not
182 changing between individuals of different ages, were assigned at the species level.

183

184 Functional space characterization

185 We performed functional analyses following the general protocol proposed in Mammola et al.
186 (2020c). We expected the properties of the functional space to vary between the two different
187 habitats, reflecting the habitat filtering effect in sorting the mite communities according to the
188 presence of certain traits. Furthermore, we expected variations in the functional space in relation
189 to the phenological changes of the *P. oceanica* meadow through the four sampling campaigns.

190 We represented the functional space of mite communities in the two habitats and across
191 sampling campaigns with geometrical *n*-dimensional hypervolumes (Blonder et al., 2014, 2018).
192 Since some of the functional traits considered here are categorical, we applied a Gower
193 dissimilarity measure to the complete trait matrix and extracted orthogonal morphological axes
194 through principal coordinate analysis (Carvalho & Cardoso, 2020; Mammola & Cardoso, 2020).
195 We delineated hypervolumes with the package '*hypervolume*' (Blonder & Harris, 2018) of the R
196 software (R Core Team, 2020) using a gaussian kernel density estimate (Blonder et al., 2014,
197 2018), the first four principal coordinate axes (cumulatively 60% variance explained), a default
198 bandwidth for each axis, and species abundances. A gaussian kernel density estimation was
199 selected as it allows a probabilistic rather than a binary characterization of the functional space
200 (Mammola & Cardoso, 2020). Five samples with one or no species were removed from the

201 analyses. We analysed the properties of the hypervolumes with specific indices (Mammola &
202 Cardoso, 2020) implemented in the R package ‘BAT’ (Cardoso et al., 2015, 2020). For each set of
203 analyses, we expressed functional diversity as the total volume of the functional space. We verified
204 if communities in matte and leaves and across sampling campaigns were subjected to different
205 filtering processes by calculating the dispersion of the functional space with the *kernel.dispersion*
206 function and the ‘divergence’ method (Mammola & Cardoso, 2020). The regularity of traits
207 distributions within the total functional space expresses evenness as the overlap between the input
208 hypervolume and a theoretical hypervolume whose traits and abundances are evenly distributed
209 within their possible range, using the *kernel.evenness* function (Mammola & Cardoso, 2020).

210 We inspected whether certain assemblages of mite species act as indicators of the two
211 habitats, and which species contribute most original traits to each habitat (i.e., functional outliers;
212 Violle et al., 2017). In particular, we expect the distribution of the originality values to have a
213 smaller variation in the leaves than in the matte, reflecting the stronger filtering effect exerted by
214 this habitat compared to the matte. We calculated the functional originality of each species in each
215 community with the function *kernel.originality*, weighting originality by species abundance
216 (Mammola & Cardoso, 2020). We expressed originality as the average distance between each
217 species to a sample of 10% stochastic points within the boundaries of the hypervolume. For each
218 habitat, we expressed the total originality of a species as the average originality of the species
219 across all communities in which it was present. Also, in this analysis, we considered the stages of
220 the same species separately.

221 To define the degree to which a given species was characteristic to one habitat or the other,
222 we further calculated the Δ Originality by subtracting to the value of originality of each species in
223 the matte the value of originality of the same species in the leaves. When a species was absent in
224 a habitat, we assigned its originality in this habitat to zero. We visualized Δ Originality values as
225 histograms centred to the value of zero, where positive values indicate species that are more
226 original in the matte than in the leaves, and negative values *vice versa*. We estimated and visualized
227 the theoretical density of values with the R package ‘ggplot2’ (Wickham, 2016), by computing a
228 kernel density estimate with a default bandwidth through the data.

229 To ease the interpretation of our findings, we finally calculated the probability of
230 recovering a given trait within each habitat as the community weighted mean with the *cwm*
231 function in ‘BAT’. For categorical traits, we calculated instead the probability of finding each state

232 of the trait in each habitat using a function developed *ad hoc* for this study—see R code uploaded
233 alongside this submission.

234

235 Statistical analyses

236 We performed analysis of variance (ANOVA) to evaluate the significance of the differences
237 observed in functional diversity, dispersion, and evenness between the matte and the leaves
238 samples (Hypothesis 1). Then, we verified whether the originality values of species in the leaves
239 were significantly different than those in the matte using a null modelling approach (Hypothesis
240 2). We performed 99 permutations of the species between the two habitats, keeping fixed the
241 original abundance values. For each run, we recalculated the hypervolumes and the originality
242 values and estimated how many species in the leaves had higher originality than the species in the
243 matte. As in Mammola et al. (2020b), the null hypothesis of random sorting of species between
244 the two habitats was rejected if the observed value was higher than the 97.5 percentile or lower
245 than the 2.5 percentile of the 99 randomizations. For each permutation, we estimated the standard
246 effect size and associated p-value.

247 In order to address Hypothesis 3, we explored the variation of functional metrics across
248 sampling campaigns within each habitat using linear models (LMs). The response variables were
249 the functional metrics richness, dispersion, and evenness calculated for the mite communities in
250 each sample. As environmental predictors, we selected four variables: two of them, the length of
251 the leaves and the organic matter content in the matte, were used as proxies of food availability in
252 each habitat; the other two, the density of leaves and the density of matte, were used as proxies of
253 habitat availability. Prior to the analyses, we checked collinearity among predictors with Pearson's
254 r correlations, setting the threshold for collinearity at $|r| > 0.7$ (Zuur et al., 2010). We log-
255 transformed each independent variable in order to capture their biological effect on the mite
256 communities, which is expected to change logarithmically, *i.e.*, a difference of 1 cm in the leave
257 length is expected to have a stronger effect on the mite communities when the leaves are short than
258 when they are long. To facilitate model convergence, we further scaled all independent variables.
259 Finally, to take into account the dependency structure in our data due to sampling campaigns, we
260 included the variable sampling campaign as a fixed factor in all the models, because we could not
261 include it as a random effect due to the presence of only four levels, which are considered too few
262 to be used as a random effect (Gelman & Hill, 2006).

263 Given that the environmental predictors are different between the matte and the leaves, we
264 fitted separate regressions for the two habitats. All analyses were performed in R. Following Zuur
265 & Ieno (2016), we validated models by checking the normality of model residuals, the plot of
266 residuals versus fitted values, normal Q-Q plots, and Cook's distances, using the R package
267 'performance' (Lüdecke et al. 2020). The outputs of the results are presented as type-II analysis-
268 of-variance tables for model objects obtained with the R package 'car' (Fox & Weisenberg, 2018).
269

270 **Results**

271 We successfully reconstructed the hypervolumes for the 43 communities (that is, all those with
272 more than one species). As we expected on our Hypothesis 1, we observed a clear polarization of
273 the trait space according to the two habitats (Figure 1). Properties of the functional space of the
274 community in the two habitats were significantly different: the communities in the matte were
275 functionally more diverse (ANOVA: $F_{(1,41)} = 26.94$, $p < 0.001$), more dispersed ($F_{(1,41)} = 20.93$, $p <$
276 0.001), and more even ($F_{(1,41)} = 74.75$, $p < 0.001$) than those in the leaves (Figure 2A, Table 3).

277 Contrary to our Hypothesis 2, the distribution of the total functional originality values was
278 similar in both habitats (Figure 3A). According to the null modelling analysis, the number of
279 species more original in the leaves than in the matte was not lower than what is expected from a
280 random sorting of species across habitats (Standard effect size = -0.41 , p -value = 0.06). Regarding
281 the values of Δ Originality, we found a set of distinct species in the two habitats, allowing us to
282 differentiate the leaves and matte communities according to the functional traits of few indicator
283 species (Figure 3B).

284 The environmental predictors for each habitat were not collinear (for leaves: length *vs.*
285 density of the leaves, $r = -0.003$; for matte: organic matter *vs.* density of the matte, $r = -0.48$) and
286 were thus retained in the statistical models. Richness, dispersion, and evenness of the mite
287 communities in the leaves were only marginally negatively affected by the length of the leaves,
288 with dispersion and evenness different between sampling campaigns (Figure 2G–I; Table 4). No
289 significant effects were detected in the matte (Table 4). These results partially support our
290 Hypothesis 3, although the effect of the environment in the leaves was nonetheless weak.

291

292 **Discussion**

293 Habitat patterns in functional diversity

294 Our analyses confirmed our first hypothesis that mite communities in matte habitat had a
295 significantly higher functional richness, dispersion, and evenness than those in the leaves.
296 Analytically, this means that, on average, the functional space in the leaves is significantly less
297 voluminous (*i.e.* trait diversity is lower) and observations are less dispersed (*i.e.* species have traits
298 that are more similar amongst them) and less even (*i.e.* the traits hypervolume is not homogenous
299 indicating that certain combinations of traits are more common than others) than in the matte.
300 Biologically, this suggests that the selective conditions in the leaves exert a stronger filtering effect
301 upon the traits present in the colonizing species, whereby only a small subset from all the pool of
302 traits present in the seagrass meadow allows mites to thrive in the leaves. This habitat filtering is
303 reflected in the distribution of mites between habitats: even if the habitats are physically connected,
304 communities in the leaves consist of a subset of the species present in the matte. Furthermore, this
305 pattern was consistent through the different sampling campaigns, despite the stark phenological
306 changes experienced by the *Posidonia* meadow throughout the year. The leaves are the habitat in
307 which it is more likely to find individuals bearing specialised traits (Supplementary Material
308 Figure S1). These traits are chiefly specialised claws (Figure S1d, S1e), which might aid in clinging
309 to the leaf's surface and thereby withstand turbulence (e.g. Pfingstl et al., 2020; but see Pugh et
310 al., 1987) and a larger body size (Figure S1g). In contrast, the assemblages in the matte consist of
311 species bearing these traits, as well as species with more slender bodies (Figure S1i) and a longer
312 and pointier gnathosoma (Figure S1j). Whereas the slender body presumably aids this species to
313 crawl in the tighter habitat spaces in the matte, as observed in most interstitial microscopic species
314 (Giere 2009), it is more difficult to interpret the functional meaning of the elongation of the
315 gnathosoma. We here speculate that it might aid this species in reaching food particles accumulated
316 in the tight spaces such as detritus and deposits of organic matter, but more in-depth studies would
317 be needed to corroborate this assumption. A third group of species, presumably consisting of
318 predators feeding on mites (Bartsch, 1989; Green & MacQuitty, 1987), are found occasionally in
319 some of the samples, occurring stochastically both in the leaves and the matte as they wander
320 around in the meadow searching for their prey.

321 This general pattern further emerges from the analysis of originality values, a metric that
322 averages the distance between each observation to a sample of stochastic points within the
323 boundaries of the hypervolume. It thereby measures how unique the position of individual

324 observations is in the trait hyperspace, as the distances are expected to increase as the species'
325 combination of traits becomes unique (Mammola & Cardoso, 2020). Therefore, we expected more
326 functionally original species in the matte, because species in the leaves need special adaptations
327 presumably to cope with turbulence and feed on specialised food sources. The same adaptations
328 are not required in the matte, where the presence of shelters and more diverse sources of food
329 might relax the filtering effect on species and traits. This might result in a more functionally
330 heterogeneous assemblage in which the probability of finding a given species is less dependent
331 upon their traits. Our results, however, did not support this assumption given that originality values
332 in the leaves did not differ significantly from those in the matte (Figure 3a). This might be the case
333 because the species with the highest values of originality—such as *Pelacarus aculeatus*, *Agauae*
334 *panopae*, *Agauopsis microrhyncha*, or *Agauae abyssorum*; Table S1—typically consisted of large
335 rare species with uncommon traits that facilitate predation upon other microscopic animals,
336 including mites (Bartsch, 1989; Green & MacQuitty, 1987). These species also occur in low
337 abundances and their distribution is scattered across the meadow, being found stochastically in
338 one habitat or the other. In fact, these species can be considered functional outliers (*sensu* Violle
339 et al., 2017) in that they take extreme values of Δ Originality (Figure 3b), as they only occur in
340 low numbers in either habitat, thus indicating that the filtering may act at another spatial or
341 temporal scale on them. However, we acknowledge that further studies on the feeding biology of
342 marine mites would be needed to fully understand the biological mechanisms behind the ecological
343 patterns we documented.

344

345 Phenological changes and functional diversity

346 Our results partially corroborate our third hypothesis, as we found weakly significant variations in
347 the functional diversity of mite communities in the leaves following the phenological changes of
348 biomass of *Posidonia oceanica*, specifically the change in the length of the leaves. These changes
349 permeate all metrics, which surprisingly were negatively affected by the length of the leaves, used
350 as a proxy for food availability.

351 The end of the summer is characterized in the Mediterranean by an increase of the rainfall
352 and primary production, which favours a rapid growth of *P. oceanica* in winter reaching a peak in
353 the biomass in the seagrass meadow in spring (Champenois & Borges, 2014). A large number of
354 epiphytes colonize the leaves, which get densely populated by diverse epiphytic communities

355 (Mabrouk et al., 2011; Piazzini et al., 2016), as they enlarge. Food resources are hence more
356 abundant in the leaves at their peak of production in spring, which might feedback positively the
357 mite populations in this habitat. However, instead of favouring an increase of functional diversity
358 driven by a higher abundance of resources, our results suggested the opposite, as they show a
359 marginally significant reduction of the functional dispersal and evenness in the leaves when the
360 leaves are longer. We speculate that the higher abundance of epiphytes might provide an advantage
361 to those mites that are better adapted to feed on them, increasing their relative abundance to other
362 species and favouring the homogenization of the trait space in the leaves. Furthermore, the basal
363 parts of long leaves are less exposed to hydrodynamics, as leaves themselves provide shelter from
364 the current towards the bottom (Folkard, 2005). This favours presence of a larger number of
365 macrofaunal organisms, such as fish and decapod juveniles, which find shelter in the leaves for
366 larger macrofaunal predators (Hovel *et al.*, 2002; Hovel & Fonseca, 2005), preying on the most
367 conspicuous and less specialized meiofaunal organism that colonize the leaves (Zupo and Stübing,
368 2010). We acknowledge that these explanations are tentative given our current data. Only further
369 functional ecological approaches will be able to address our hypotheses, obtaining a more holistic
370 picture of ecosystem functioning.

371 In contrast, the matte does not experience similar pronounced phenological changes and
372 we can speculate that this is the reason for which no significant changes were observed in the
373 functional diversity of mite communities in the matte.

374

375 **Conclusions**

376 Being the first study using hypervolumes to define functional properties of meiofauna
377 communities, our study highlights a potential role of the environment in affecting the distribution
378 of microscopic animals between connected habitats by filtering them according to the presence of
379 certain traits. Remarkably, this filtering effect was relatively weak, as most species were found in
380 both habitats and the filtering was mostly reflected by their relative abundances. One may argue
381 that our results of filtering effects between connected habitats might not be applied to all
382 microscopic animals more widely and that mites in seagrass meadows might represent only a
383 specific case. Habitat filtering effects might be even more subtle in other microscopic animal
384 groups, especially the soft-bodied ones, for which the functional interpretation of morphological
385 traits is often obscure and trait measurements subjected to strong artefacts due to post-mortem

386 contraction, fixation, and other bias (Higgins & Thiel, 1988). Furthermore, most microscopic
387 animals have a high probability to be passively dispersed to suboptimal habitats (Armonies, 1988;
388 Hagerman & Rieger, 1981; Hauspie & Polk, 1973), increasing the uncertainty associated with
389 habitat characterization at a small scale relevant for their biology, thus overestimating both their
390 functional and realized niches. Interestingly, our results add an extra value to the *Posidonia*
391 *oceanica* meadows: on top of their indisputable importance as a reservoir of biological diversity
392 (e.g., Mazzella & Spinoccia, 1992; Kalogirou et al. 2010; Urra et al., 2013; Piazzini et al., 2016) and
393 the many services that they provide (Boudouresque et al., 2017; Vacchi et al., 2017), they may also
394 represent important model systems to explore research questions in ecology and evolution, such
395 as distribution patterns of microscopic fauna.

396 It is not surprising that in studies on the distribution of microscopic animals, such
397 distribution might appear either uniform or random, simply as a consequence of the high
398 uncertainty associated with measurements and morphological interpretation at the small spatial
399 scales. In other words, microscopic size may generate uncertainty in a macroscopic observer, on
400 both the definition of traits and the definition of niche even if *the environment did select*. Exploring
401 the distribution of small animals through the lens of functional ecology, targeting traits with clear
402 functional meaning related to habitat occupation, is crucial to overcome some of these biases
403 (Violle et al., 2014). Our study therefore emphasises the need of moving from a merely
404 taxonomical toward a functional view of ecological studies of microscopic organisms (Green et
405 al., 2008). Further steps in this direction will warrant a better mechanistic understanding of their
406 habitat and distribution patterns.

407

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409

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422

423 AUTHOR CONTRIBUTION STATEMENTS

424

425 AM, GGG, AGH, and NS planned the sampling design. GGG, AGH, NS and AIM collected the
426 samples; GGG, AGH and NS sorted the latter samples for animals of interest and measured the
427 environmental variables, whereas GGG identified animals and collected traits. AM and SM
428 planned the statistical approach and performed analyses. FP provided facilities and support. AM,
429 GGG, and SM wrote the first draft. All authors contributed to the writing to additions and
430 comments to the text.

431

432

433 DATA AVAILABILITY STATEMENTS

434

435 Raw data and R script to generate the analyses will be deposited in a public repository upon
436 acceptance.

437

438

439 SUPPLEMENTARY MATERIALS ONLINE

440

441 **Figure S1.** Probability of finding each state of discrete traits (**a–f**) and community weighted mean
442 of continuous traits (**g–k**) for mite communities in the leaves and matte.

443

444

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649 **TABLES & FIGURES**

650 **Table 1.** Morphological traits considered in the analyses, with hypotheses on their functional
 651 meaning

Trait	Variable description	Functional meaning
(1) Total length	Measurement the tip of the gnathosoma to the tip of the idiosome in mm	Proxy of the total biovolume, trophic level and passive resistance of mites against water currents.
(2) Idiosome length	Idiosome dorsal length	Proxy of the hard body length.
(3) Idiosome width	Idiosome dorsal width	Proxy of the hard body width.
(4) Gnathosoma (dorsal) length	Length of the gnathosoma which is not covered by the idiosome and exposed dorsally.	Proxy of the diet. The length of the gnathosoma is adapted to exploit different food resources (Bartsch 2006).
(5) Idiosome length/width	Ratio between idiosome length and width	Proxy of body shape. Wider body shapes limit the colonization of habitat consisting of narrow spaces. Indeed, slender shaped mites are often found amongst fine sediments (Bartsch 2006).
(6) Relative gnathosoma length	Ratio between gnathosoma dorsal length total body length	Proxy of the diet, as a measure of protruding gnathosoma relative to body size.
(7) Number of Accessory teeth	Number of accessory teeth on the claws	In mites, especially those species linked to aquatic habitats, claws are essential to withstand physical stress, whether large (Pfungstl et al. 2020) or structural complex claws (Pugh & Fordy, 1987; Bartsch 2006). We here include four claw
(8) Combs	Degree of comb complexity, where 0 = absence, 1 = fine, 2 = regular, and 3 = large combs	

(9) Median claw type	Degree of median claw development, where 0 = absence, 1 = small, and 2 = large median claw	median claw structures to account for different possible combinations that define claw complexity. The combination of these variables provides a proxy of the
(10) Number of legs with combs	Number of pairs of legs whose claws bear combs	resistance of each individual to turbulence, as increasing claw complexity means a better grip to the substrate.
(11) Lamella	Categorical, reflecting the presence/absence of cerotegumental or cuticular lamella on legs	Lamella are present mostly in species that occur in sediments (Bartsch 2006).
(12) Pincer	Categorical, reflecting the presence of a first pair of legs modified as a pincer	Specialised legs for feeding (Green & Macquitty 1987; Bartsch 2006).

653 **Table 2.** Summary of the species included in this study, number of counted individuals, and coding for the 12 included functional
654 traits in each developmental state (\pm standard error). The number for each character and the explanation for the coding are summarized
655 in Table 1. Abbreviations: N, total number of measured specimens.
656

Species name	N	stage	1	2	3	4	5	6	7	8	9	10	11	12
<i>Agauae cf. abyssorum</i> (Trouessart, 1896)	1	juvenile	394.58 \pm NA	323.52 \pm NA	242.92 \pm NA	71.06 \pm NA	1.33 \pm NA	0.18 \pm NA	0	1	0	4	1	0
<i>Agauae panopae</i> (Lohmann, 1893)	7	adult	737.42 \pm 61.05	587.59 \pm 52.18	366.99 \pm 48.21	149.83 \pm 17.50	1.61 \pm 0.07	0.20 \pm 0.02	1	2	1	4	1	0
	5	juvenile	621.77 \pm 139.82	480.90 \pm 111.47	314.33 \pm 91.55	140.87 \pm 39.26	1.55 \pm 0.12	0.23 \pm 0.04	1	2	1	4	1	0
<i>Agauopsis brevipalpus</i> (Trouessart, 1889)	3	adult	483.66 \pm 24.32	389.34 \pm 4.78	294.59 \pm 17.52	94.32 \pm 25.56	1.33 \pm 0.09	0.19 \pm 0.04	1	2	0	3	0	0
	3	juvenile	366.14 \pm 52.34	301.28 \pm 45.14	231.84 \pm 41.69	64.86 \pm 7.58	1.30 \pm 0.06	0.18 \pm 0.01	1	2	0	3	0	0
<i>Agauopsis microrhyncha</i> (Trouessart, 1889)	5	adult	520.87 \pm 15.70	487.68 \pm 9.70	342.02 \pm 27.43	40.13 \pm 10.61	1.38 \pm 0.02	0.08 \pm 0.02	1	3	1	4	0	0
	9	juvenile	376.29 \pm 71.24	321.94 \pm 39.12	236.68 \pm 30.76	33.75 \pm 7.61	1.37 \pm 0.10	0.10 \pm 0.03	1	3	1	4	0	0
<i>Agauopsis minor</i> (Trouessart, 1894)	10	adult	377.19 \pm 10.33	340.37 \pm 13.98	245.66 \pm 10.81	36.82 \pm 7.47	1.39 \pm 0.05	0.10 \pm 0.02	1	2	1	3	0	0
	4	juvenile	267.27 \pm 82.22	237.20 \pm 70.45	170.28 \pm 58.76	30.07 \pm 15.04	1.41 \pm 0.09	0.11 \pm 0.03	1	2	1	3	0	0
<i>Arhodeoporus gracilipes</i> (Trouessart, 1889)	10	adult	353.28 \pm 19.15	291.01 \pm 16.15	190.17 \pm 19.34	62.28 \pm 7.48	1.57 \pm 0.09	0.19 \pm 0.03	1	1	1	4	0	0
	5	juvenile	277.26 \pm 48.32	223.23 \pm 36.39	145.35 \pm 35.27	54.03 \pm 13.32	1.56 \pm 0.14	0.19 \pm 0.02	1	1	1	4	0	0

<i>Arhodeoporus labronicus</i> (Morselli, 1981)	1	adult	303.75 ± NA	245.66 ± NA	130.81 ± NA	58.09 ± NA	1.88 ± NA	0.19 ± NA	1	1	1	3	0	0
	2	juvenile	281.05 ± 33.72	231.54 ± 33.71	132.52 ± 35.11	49.51 ± 0.01	1.78 ± 0.22	0.18 ± 0.02	1	1	1	3	0	0
<i>Copidognathus lamelloides</i> Bartsch, 2000	24	adult	337.40 ± 16.96	280.95 ± 14.69	201.56 ± 12.92	56.44 ± 6.82	1.40 ± 0.05	0.17 ± 0.02	1	1	1	3	1	0
	11	juvenile	242.00 ± 35.00	196.23 ± 32.85	129.10 ± 21.39	45.77 ± 3.84	1.52 ± 0.05	0.19 ± 0.02	1	1	1	3	1	0
<i>Copidognathus latisetus</i> Viets, 1940	15	adult	219.34 ± 6.72	203.85 ± 6.11	122.41 ± 6.42	15.49 ± 3.72	1.68 ± 0.08	0.07 ± 0.02	1	1	1	4	0	0
<i>Copidognathus magnipalpus</i> (Police, 1909)	21	adult	398.95 ± 17.49	339.31 ± 18.23	220.12 ± 17.73	59.64 ± 6.16	1.55 ± 0.08	0.15 ± 0.02	2	3	1	4	0	0
	8	juvenile	291.29 ± 51.78	254.38 ± 48.83	170.00 ± 24.15	46.63 ± 8.59	1.42 ± 0.26	0.16 ± 0.01	2	3	1	4	0	0
<i>Copidognathus oculatus</i> (Hodge, 1863)	30	adult	352.82 ± 14.48	299.28 ± 13.41	176.19 ± 12.48	53.54 ± 9.01	1.71 ± 0.10	0.15 ± 0.02	1	3	1	4	0	0
<i>Copidognathus quadricostatus</i> (Trouessart, 1894)	1	adult	382.72 ± NA	301.41 ± NA	212.16 ± NA	81.31 ± NA	1.42 ± NA	0.21 ± NA	1	1	1	4	0	0
	1	juvenile	249.17 ± NA	198.32 ± NA	113.43 ± NA	50.85 ± NA	1.75 ± NA	0.20 ± NA	1	1	1	4	0	0
<i>Copidognathus remipes</i> (Trouessart, 1894)	13	adult	360.86 ± 10.87	299.25 ± 11.24	175.32 ± 11.91	60.45 ± 7.05	1.71 ± 0.13	0.17 ± 0.02	1	0	1	4	1	0

	5	juvenile	301.05 ± 70.17	248.63 ± 60.69	153.42 ± 37.26	52.42 ± 10.87	1.62 ± 0.09	0.18 ± 0.02	1	0	1	4	1	0
<i>Copidognathus reticulatus</i> (Trouessart, 1893)	1	juvenile	269.36 ± NA	225.19 ± NA	124.03 ± NA	44.17 ± NA	1.82 ± NA	0.16 ± NA	1	1	1	4	1	0
<i>Lohmannella falcata</i> (Hodge, 1863)	5	adult	494.44 ± 28.73	326.74 ± 15.22	257.21 ± 18.69	167.70 ± 19.75	1.27 ± 0.06	0.34 ± 0.02	0	0	0	0	0	0
	3	juvenile	304.25 ± 57.12	210.79 ± 51.87	166.02 ± 41.00	93.46 ± 0.02	1.27 ± 7.64	0.31 ± 0.04	0	0	0	0	0	0
<i>Pelacarus aculeatus</i> (Trouessart, 1896)	2	adult	574.82 ± 118.04	479.31 ± 95.03	401.39 ± 98.56	95.51 ± 23.01	1.20 ± 0.06	0.17 ± 0.01	2	0	1	0	0	0
	2	juvenile	389.57 ± 0.61	313.10 ± 5.27	243.75 ± 10.87	76.48 ± 4.66	1.29 ± 0.04	0.20 ± 0.01	2	0	1	0	0	0
<i>Rhombognathus praegracilis</i> Viets, 1939	110	adult	398.11 ± 26.82	348.55 ± 25.75	237.16 ± 22.80	50.28 ± 8.13	1.48 ± 0.11	0.13 ± 0.02	2	3	0	4	0	0
	172	juvenile	289.88 ± 60.37	246.25 ± 53.94	166.26 ± 37.81	34.84 ± 7.98	1.51 ± 0.10	0.13 ± 0.03	2	3	0	4	0	0
<i>Rhombognathus cf. procerus</i> Bartsch, 1975	1	adult	333.83 ± NA	297.33 ± NA	198.00 ± NA	36.50 ± NA	1.50 ± NA	0.11 ± NA	1	1	0	4	0	0
<i>Simognathus minutus</i> (Hodge, 1863)	5	adult	450.79 ± 20.79	374.39 ± 19.23	202.92 ± 43.67	76.40 ± 6.40	1.93 ± 0.54	0.17 ± 0.01	1	1	2	3	0	1
	7	juvenile	361.95 ± 45.32	298.18 ± 41.61	197.88 ± 41.56	53.03 ± 5.00	1.59 ± 0.13	0.16 ± 0.02	1	1	2	3	0	1

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Table 3. Summary of the average values (\pm standard error) of the number of species, number of individuals, and hypervolume metrics for the samples grouped by habitat (leaves and matte) and sampling campaign.

Habitat	Sampling campaign	Richness	Dispersion	Evenness	Number of species	Number of individuals
leaves	total	0.007 \pm 0.002	0.204 \pm 0.009	0.076 \pm 0.011	6.792 \pm 0.481	58.583 \pm 13.127
	December	0.001 \pm 0.000	0.159 \pm 0.005	0.029 \pm 0.015	6.667 \pm 0.615	146.167 \pm 31.584
	March	0.011 \pm 0.004	0.225 \pm 0.017	0.105 \pm 0.018	6.333 \pm 1.202	24.333 \pm 3.148
	April	0.014 \pm 0.004	0.247 \pm 0.012	0.124 \pm 0.017	7.000 \pm 1.033	22.167 \pm 2.701
	August	0.003 \pm 0.001	0.185 \pm 0.013	0.046 \pm 0.015	7.167 \pm 1.138	41.667 \pm 8.053
matte	total	0.026 \pm 0.004	0.261 \pm 0.008	0.213 \pm 0.011	8.000 \pm 0.662	15.053 \pm 1.822
	December	0.025 \pm 0.004	0.262 \pm 0.013	0.216 \pm 0.023	6.600 \pm 1.364	13.2000 \pm 3.967
	March	0.019 \pm 0.005	0.243 \pm 0.016	0.189 \pm 0.016	9.400 \pm 1.833	20.200 \pm 5.305
	April	0.036 \pm 0.008	0.285 \pm 0.009	0.239 \pm 0.022	7.667 \pm 0.803	13.000 \pm 1.592
	August	0.021 \pm 0.01	0.239 \pm 0.027	0.194 \pm 0.012	8.667 \pm 0.882	13.667 \pm 0.333

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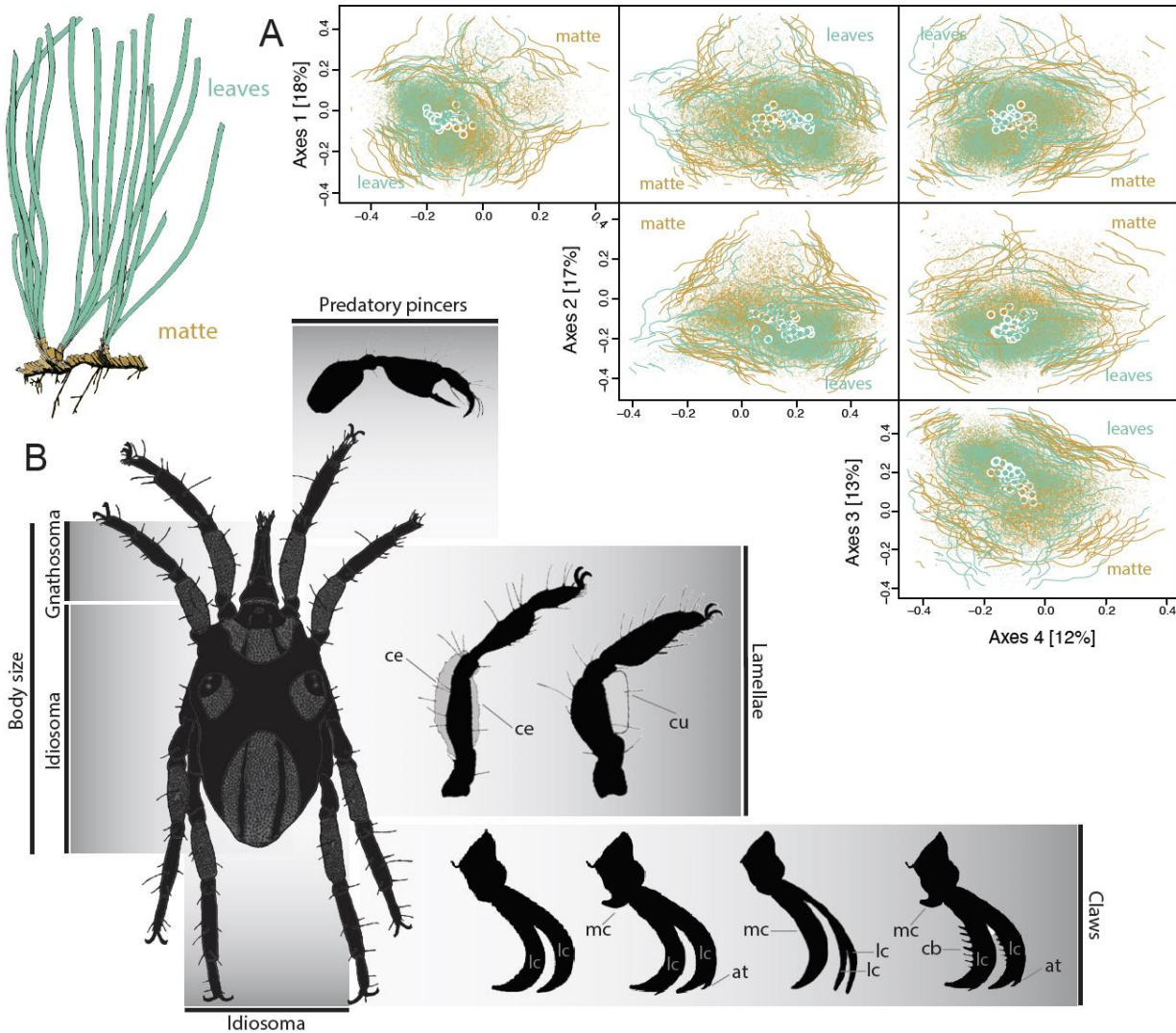
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668 **Table 4.** Results of the linear models between metrics of functional diversity and ecological
669 variables, reported as type-II analysis-of-variance tables. Continuous predictors are log-
670 transformed. Bold values denote significant effects. Abbreviation: df = degrees of freedom
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Habitats	Response variables	Environmental predictors	df	F -value	p-value
Leaves	Richness	density of leaves	1	0.409	0.530
		length of leaves	1	4.543	0.047
		sampling campaign	3	2.980	0.059
		residuals	18		
	Dispersion	density of leaves	1	0.268	0.611
		length of leaves	1	4.667	0.044
		sampling campaign	3	5.368	0.008
		residuals	18		
	Evenness	density of leaves	1	0.001	0.976
		length of leaves	1	5.325	0.033
		sampling campaign	3	4.681	0.014
		residuals	18		
Matte	Richness	density of matte	1	0.007	0.937
		organic of matter	1	2.416	0.144
		sampling campaign	3	1.256	0.330
		residuals	13		
	Dispersion	density of matte	1	0.392	0.542
		organic of matter	1	1.268	0.280
		sampling campaign	3	1.856	0.187
		residuals	13		
	Evenness	density of matte	1	0.391	0.543
		organic of matter	1	0.026	0.875
		sampling campaign	3	0.676	0.582
		residuals	13		

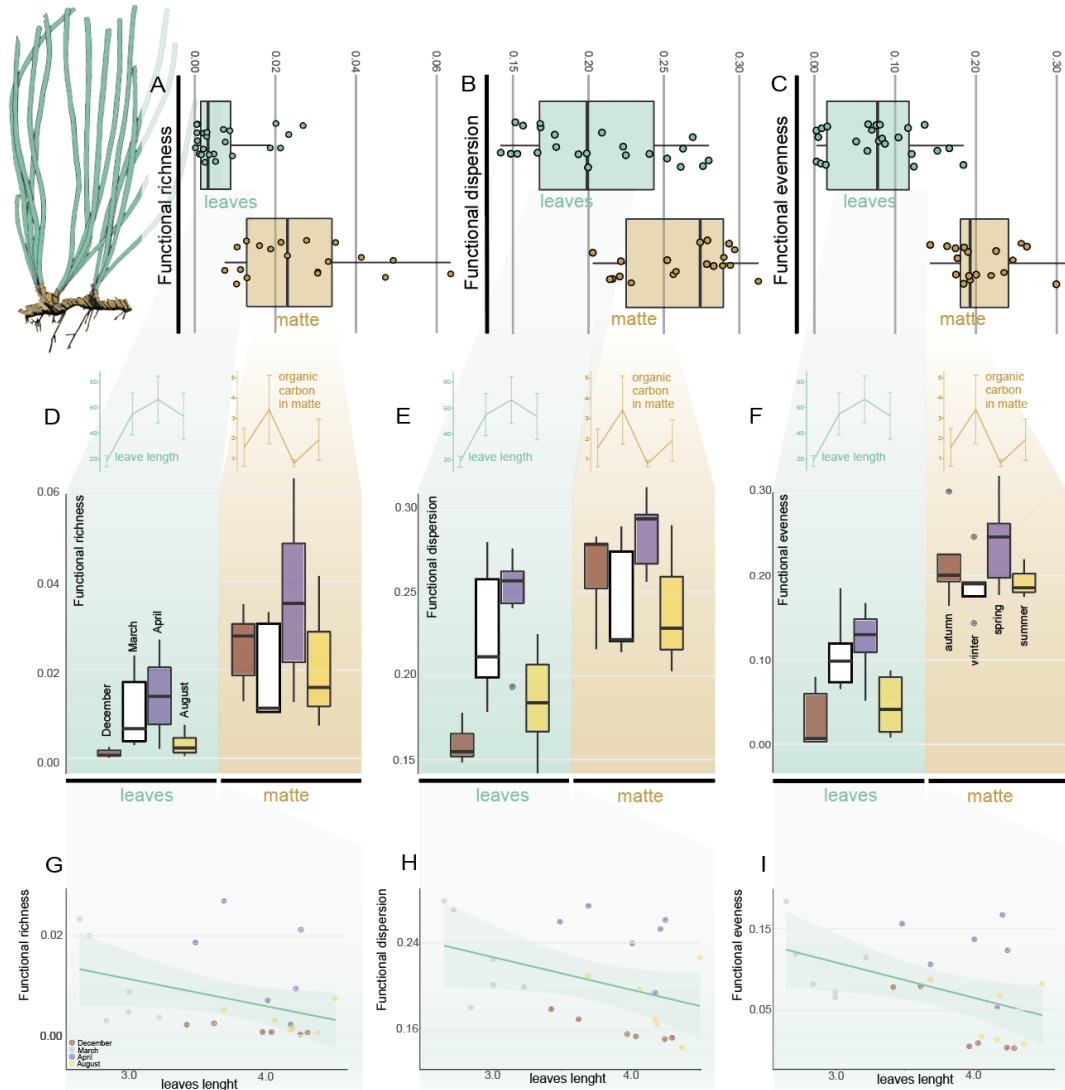
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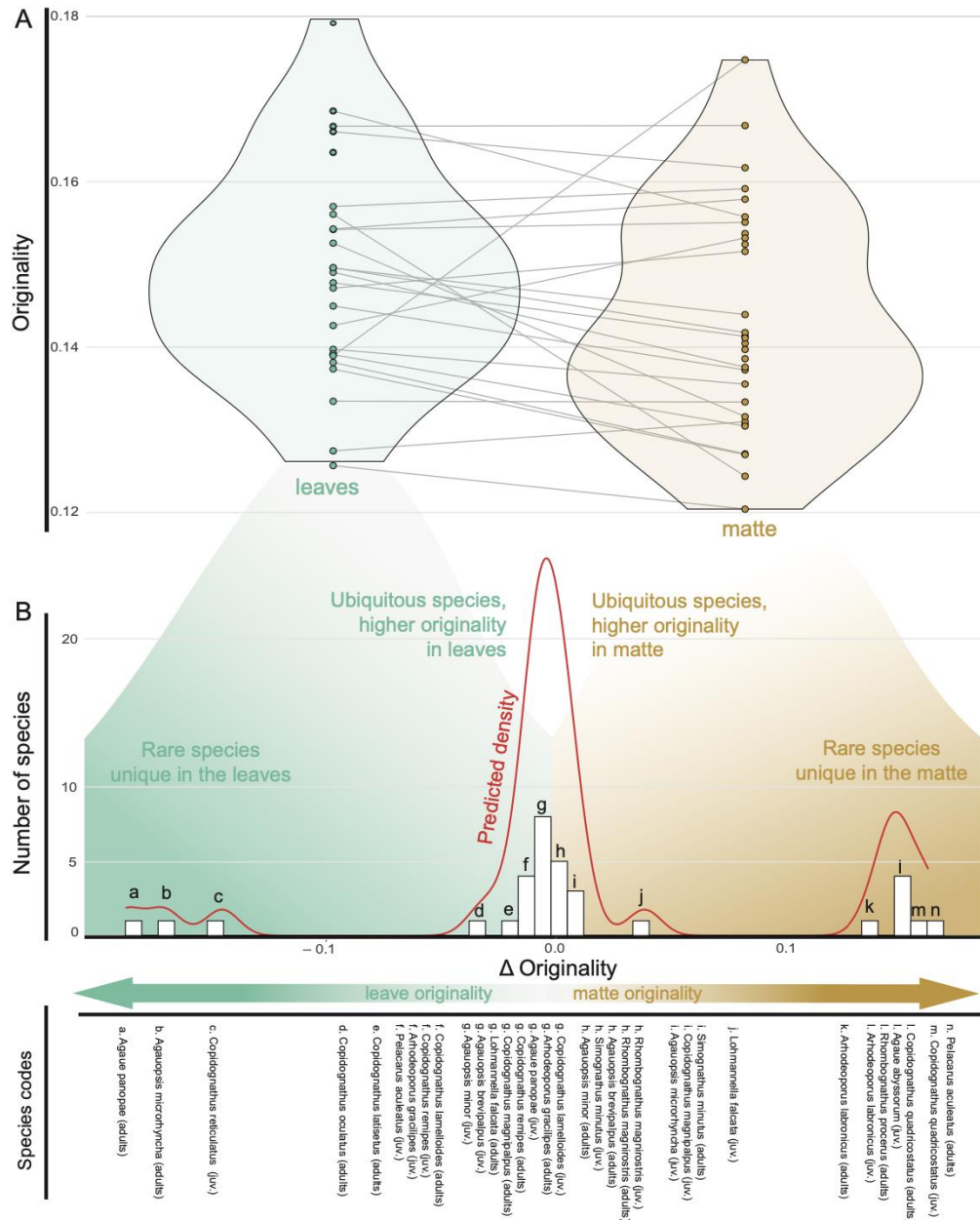


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 675 **Figure 1.** A) The 4-dimensional hypervolume of the mite communities in the *Posidonia oceanica*
 676 leaves (n=24) and matte (n=19). Large points with white borders represent the centroid of each
 677 hypervolume (note that due to the proximity of centroids, most points appear superimposed). The
 678 shape and boundaries of each hypervolume are defined by 1,000 random points. All points are
 679 coloured according to the habitat. B) Summary of the morphological traits measured or estimated
 680 for each species and developmental stage. Further details on the interpretation of each trait are
 681 provided in Table 1 and 2, and the average values of traits across habitats in Figure S1.
 682 Abbreviations: *at* accessory tooth, *cb* comb, *ce* ceratogegumental lamellae, *cu* cuticular lamellae,
 683 *lc* lateral claw, *mc* median claw.

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 686 **Figure 2.** A–C) Overall differences in functional richness (A), dispersion (B) and evenness (C)
 687 between mite communities in leaves and matte. D–F) Differences in functional richness (D),
 688 dispersion (E) and evenness (F) across sampling campaigns. Each sampling campaign corresponds
 689 to a different period along the full phenological cycle of *Posidonia oceanica*. Inset graphs in D–F
 690 represent the variation in leaves mean length (in cm) for the leaves, and the organic matter content
 691 (in %) for the matte, thus reflecting the change in energy inputs due to the regeneration of leaves
 692 in the seagrass meadow across the four sampling campaigns. G–I) Effect of leaves length on
 693 functional richness (G), dispersion (H), and evenness (I); the regression lines together with the
 694 95% confidence intervals are reported, and colours of the dots refer to the four sampling
 695 campaigns.



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697 **Figure 3. A)** Violin plots showing the distribution of functional originality values of species in the

698 leaves and the matte. Grey lines connect species that are present in both habitats. **B)** Histogram of

699 Δ Originality values between species in the two habitats, calculated by subtracting the value of

700 originality of each species in the leaves to the value of originality of each species in the matte.

701 Orange smoothed lines show the predicted density of values according to a kernel density

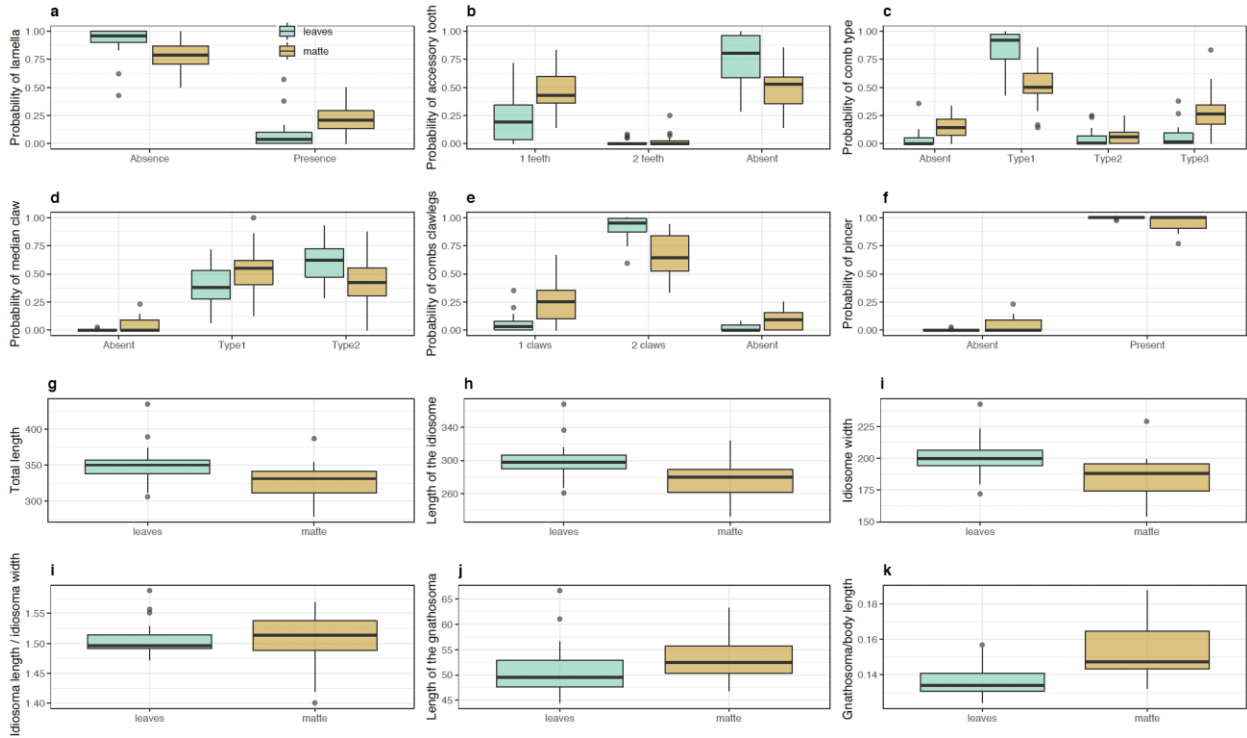
702 estimation. The letters above each bar correspond to the species listed at the bottom of the figure.

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Supplementary material Figure S1

Habitat differences filter functional diversity of low dispersive microscopic animals



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Figure S1. Probability of finding each state of discrete traits (a–f) and community weighted mean of continuous traits (g–k) for mite communities in the leaves and matle.



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Supplementary Material
02_Supplementary_table.xlsx

