

1 CAN GEOCACHING BE AN INDICATOR OF CULTURAL ECOSYSTEM
2 SERVICES? THE CASE OF THE MONTADO SAVANNAH-LIKE
3 LANDSCAPE

4 **ACCEPTED VERSION**

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38 **KEYWORDS:**
39 Crowdsourcing databases, land use, spatial analysis, stated and revealed preferences, Iberian
40 Peninsula

41
42 **Highlights:** Please see also the attached file.
43 Geocachers did not show a priori preferences for different types of land use
44 After the visit, their stated preferences exposed the appreciation of the *montado* cultural landscape over
45 other forest landscapes; they also expressed preferences for open and aquatic landscapes
46 Geocaching is a good indicator for cultural ecosystem services
47 *Montado* landowners can diversify activities on their properties, decreasing their dependence of
48 provisioning services

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53 **Abstract**

54 Cultural Ecosystem Services (CES) are difficult to assess and are seldom considered by land
55 managers. Geocaching, an outdoor game that uses Global Positioning System (GPS) enabled
56 devices to find hidden containers (geocaches) in certain locations, has been seldom used as a
57 data source to assess CES. However, contrary to other crowdsourcing databases, geocaching
58 allows to associate particular experiences to accurate locations. Furthermore, databases
59 generated by geocachers provide an ideal case to compare revealed preferences (the frequency
60 of visits to a specific geocache) with stated preferences (*a posteriori* evaluation of each location).
61 We tested the relevance of geocaching databases as CES indicators using a dataset of 50 818
62 geocaches spread across continental Portugal, over eight land-use classes, with a focus on the
63 *montado* (a high nature value farmland found in Southwestern Iberian Peninsula). We found that
64 site visitation frequency was related with its availability, showing no revealed preference towards
65 any land use. However, site evaluations by geocachers, measured either through the number
66 of words describing the experience, the number of photos taken, or the number of votes for
67 “favourite geocaches”, showed marked differences in their stated preferences, with higher
68 appreciation for open land uses in general, and *montado* in particular, especially when
69 compared to other forested landscapes. Our results may contribute to the design of regional
70 development and land-use management policies of this threatened landscape, since they show
71 the system’s strong potential as CES provider and, consequently, promoter of diversification of
72 activities.

73

74 **1. Introduction**

75 Cultural Ecosystem Services (CES) are defined as “the nonmaterial benefits people obtain from
76 ecosystems through spiritual enrichment, cognitive development, reflection, recreation, and
77 aesthetic experiences” (Millennium Ecosystem Assessment, 2005). This kind of interaction
78 between people and nature implies an emotional connection and the creation of strong ties with
79 the landscapes, nourishing the feeling of being “at home” (Schaich et al., 2010). The value of
80 CES among all Ecosystem Services (ES) is high, either in more industrialized societies, where
81 CES have an expected tendency to increase in importance, or in more traditional communities,
82 where they are often essential for cultural identity and even survival (Milcu et al., 2013). For
83 people in developed countries CES represent one of the strongest incentives to become
84 involved in environmental conservation (Phillips, 1998), and this tendency is also becoming
85 more noticeable in less developed regions (Sodhi et al., 2010).

86 CES are nowadays considered a fundamental component of the ES frameworks, often
87 influencing more the acquisition and management of the land than traditional commodity
88 production (Bieling, 2004; Plieninger et al., 2012). On the other hand, CES have less potential

89 for mediation by socioeconomic factors, as compared to other ES, which means that, once
90 degraded they are unlikely to be replaced by technical or other means (MA, 2005).

91 Cultural landscapes are defined as “landscapes with a long history, which evolved slowly and
92 where it took centuries to form a characteristic structure reflecting a harmonious integration of
93 abiotic, biotic and cultural elements” (Antrop, 1997). These landscapes have the peculiarity of
94 having an added value, as compared to more “natural” habitats, since besides the natural value,
95 they hold cultural value, due to the long and complex history of coexistence with man (Schaich
96 et al., 2010). However, cultural landscapes are undergoing rapid transformations across the
97 world, driven by abandonment and rural exodus on the one hand, and intensification of human
98 use on the other (Plieninger et al., 2014).

99 The Portuguese *montado* is a cultural landscape subject to the same type of rapid
100 transformation, affected by several threats and drivers of change, such as abandonment
101 (Bugalho et al., 2011; Godinho et al., 2016), tree mortality (Costa et al., 2010) overgrazing
102 (Almeida et al., 2015; Gonçalves et al., 2012), mechanised ploughing (Pinto-Correia, Ribeiro, &
103 Sá-Sousa, 2011) or climate change (Correia et al., 2018; López-Tirado et al., 2018; Ogaya and
104 Peñuelas, 2006; Vessella et al., 2017).

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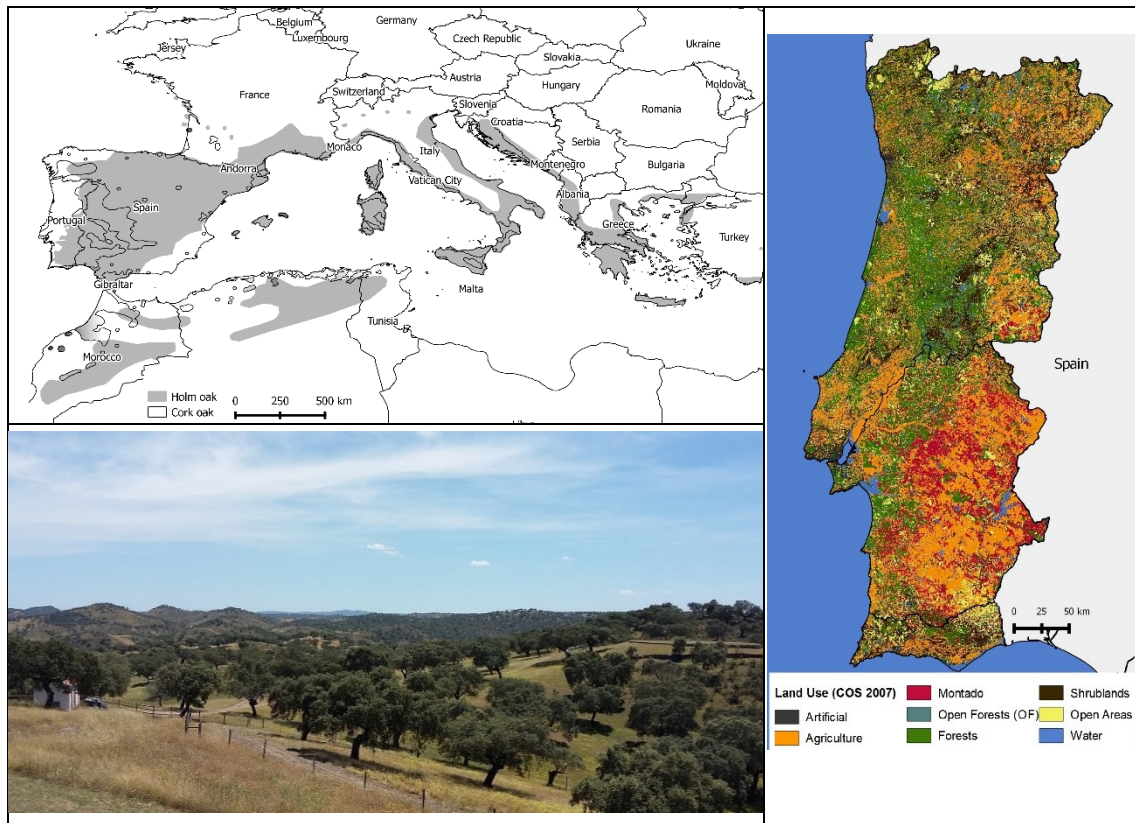
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110 *Montados* are agro-silvo-pastoral systems that resulted from millennia of traditional land use
 111 practices on the original Mediterranean woods dominated by cork oaks (*Quercus suber*) and
 112 holm oaks (*Quercus ilex* or *Q. rotundifolia*) (Joffre et al., 1988, 1999). These man-modelled
 113 ecosystems have a savannah-like aspect (Figure 1) and are characterized by scattered trees,
 114 with herbaceous understory and near absence of the shrub layer (Joffre et al., 1999).



115 **Figure 1 – Up left) distribution of Mediterranean oaks (cork oak and holm oak) adapted from Caudullo et al., 2017; bottom**
 116 **left) *Montado* landscape in southeast Portugal (Herdade da Coitadinha); right) land use map of mainland Portugal (COS –**
 117 **Carta de Uso e Ocupação do Solo- 2007 reclassified to 8 land use classes).**

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119 On *montados*, three main rural activities are pursued simultaneously within a single space:
 120 harvesting of forest products, livestock husbandry, and agriculture (Blondel, 2006). The largest
 121 cork-oak *montado* area in the world is found in southern Portugal (Alentejo region, a district
 122 inherently linked to *montado* landscape), where about 90% of the Portuguese distribution of this
 123 system is located (Figure 1). Portuguese holm oak *montados* are also more frequent in the
 124 Alentejo region and continue into Spain (where they are called *dehesas*), being part of the
 125 largest holm oak *montado* landscape in the world (Ribeiro et al., 2010). Although created and
 126 permanently managed by man, this ecosystem maintains high biodiversity levels (Blondel, 2006;
 127 Bugalho et al., 2011; Carrete and Donazar, 2005; Díaz-Villa et al., 2003; Martins da Silva et al.,
 128 2008; Pereira, 2012; Tellería, 2001), and is considered a priority habitat for conservation (Annex
 129 I of the European Union Habitats Directive 92/43/CEE) and a High Nature Value Farmland
 130 (HNVF) (EEA, 2004).

131 This multi-use system is responsible for the delivery of many ES, such as, among the regulation
132 and maintenance ES, the protection of soil against erosion by vegetation, or climate regulation
133 (Barredo et al., 2015; Guerra et al., 2016). The most recognised ES is a provision ES, namely
134 the production of cork, an important revenue for Portugal, representing 1.2% of Portuguese
135 exports (APCOR, 2016). Other relevant provisioning services are food products from Alentejo,
136 which lead the Portuguese DOP (Protected Designation of Origin) markets (Tibério and
137 Francisco, 2012). Traditional products include a variety of sheep or goat cheeses, and especially
138 the ham or sausages made from the meat of free-ranging Iberian pigs fed on holm oak acorns.
139 Besides its important market value, these products have an additional cultural value, as the clear
140 connection with the landscape provides them a regional identity. Picking up asparagus or
141 mushrooms (picked up for food by local people for centuries) are also examples of activities that
142 nowadays represent more cultural services than provisioning ones (Molina et al., 2012). Other
143 leisure activities emerged more recently through well-established events associated to this
144 landscape, e.g. the running trail “*Montado Running*” or the mountain biking “*BTT Terras do*
145 *Montado*”. Even the harvesting of the bark of the cork oak tree has a cultural value, since it is a
146 traditional hand-made activity, carried by specialised workers using only a small axe, and
147 transmitted across generations (APCOR, n.d.). People also identify this landscape with
148 traditional music (“Cante Alentejano” was classified as Intangible Cultural Heritage of Humanity
149 in 2014), traditional clothes and language (very strong and peculiar regional accent). Finally, the
150 intrinsic natural characteristics of *montado* potentiates the delivery of CES similar to other
151 natural areas, namely nature-based recreation such as hunting, fishing, birdwatching or
152 geocaching (Belo et al., 2009).

153 Measuring CES has been one of the most difficult and least accomplished tasks in ES research
154 (Millennium Ecosystem Assessment, 2005), up to today (Daniel et al., 2012; Yoshimura and
155 Hiura, 2017). The intangibility of these services is often considered the reason for their poor
156 appraisal (Hernández-Morcillo et al., 2013; Schaich et al., 2010), and CES are seldom integrated
157 in management plans and reflected in economic indicators (Milcu et al., 2013). This
158 underrepresentation results in biased ES assessments (Hernández-Morcillo et al., 2013) and
159 management planning, preventing the integration of CES into policy measures.

160 Some researchers defend that CES cannot be split into discrete units for marginal valuation
161 (Abson and Termansen, 2011), and there are suggestions for the use of indicators attributed to
162 particular landscape characteristics which hold several CES (Norton et al., 2012; Plieninger and
163 Bieling, 2012). Participatory mapping (Fagerholm et al., 2012; Plieninger et al., 2013b; Van
164 Berkel and Verburg, 2014) or photo-based methods (Almeida et al., 2014; Oteros-Rozas et al.,
165 2017) are among the methods used to define consensual indicators. More recently,
166 crowdsourcing indicators available in social networks such as Panoramio, Flickr, Facebook,
167 Twitter or Instagram, are becoming popular (Figuerola-Alfaro and Tang, 2017; Gliozzo et al.,

168 2016; Tenerelli et al., 2016; Yoshimura and Hiura, 2017). These indicators are mostly estimated
169 from geotagged photographs voluntarily uploaded by users in social networks in high numbers.
170 Similarly to the social networks mentioned above, geocaching can provide visual and written
171 information through its online platforms, but studies using these data to assess CES are very
172 scarce and, to our knowledge, only related to recreational services (Cord et al., 2015; Mendes
173 et al., 2013; Santos et al., 2014, 2012).

174 Geocaching is an outdoor game where users (geocachers) use Global Positioning System
175 (GPS) enabled devices to find hidden containers, known as geocaches (or simply caches), and
176 then sign a logbook to record and share the visit with the geocacher community. According to
177 the game rules, each attempt to find a cache must also be registered (logged) on the geocache
178 web-page, on the official geocaching website, by writing the geocacher experience, uploading
179 photos, and voting on favourite caches (see methods for more information on this). The
180 availability of exact locations and additional information, comparatively to other online platforms,
181 is an opportunity to explore the feasibility of using geocachers' logs as CES indicators, since,
182 while playing this game, geocachers can benefit from several bundles of CES, such as
183 aesthetics, bequest, recreation or inspiration for art. Motivation to find a particular geocache can
184 derive from many factors (Cord et al., 2015; O'Hara, 2008; Vartiainen and Tuunanen, 2013),
185 and not all of them are related to the landscape. However, regardless of his(her) motivations,
186 throughout the journey and culminating at the geocache, each geocacher will necessarily
187 experience the landscape; this interaction can be stated in the geocacher's log and become a
188 source of information.

189 In this study we aim to demonstrate the usefulness of the until now largely unexplored
190 geocaching databases as CES indicators and focused our analysis on the *montado* cultural
191 landscape. With an innovative approach, we used both the number of visits to each geocache
192 and the amount of information stated in the logs (whether text, photos or votes) to evaluate
193 geocachers emotional connection with the different land use classes. First, to detect a potential
194 *a priori* drive towards a particular type of landscape, we compared the frequency of visits to
195 geocaches located in different land uses with their availability. Second, we assessed the stated
196 experience of geocachers at each land use by comparing the number of words, photos or votes
197 per log. We assumed that the time invested in writing or taking pictures (translated into number
198 of words or photos, respectively) should be proportional to the intensity of the emotion felt by
199 geocachers. Given the human preference for open landscapes that allow a view over wide areas
200 (Hartmann and Apaolaza-Ibáñez, 2010; Kaplan, 1987; Mendes et al., 2014), we hypothesized
201 that the savannah-like *montado* should be preferred over closed forest landscapes. We also
202 hypothesized that variables affecting the sense of vastness, such as altitude and landscape
203 heterogeneity, should be influential on the geocachers' experience.

204 **2. Materials and Methods**

205 **2.1. Study area**

206 We considered mainland Portugal as our study area, covering a roughly rectangular 89.060 km²
207 area in the Western Iberian Peninsula (Figure 1). Higher altitudes occur in the north, reaching
208 1991m in Serra da Estrela, while through the south of the Tagus River flat plains are dominant.
209 Seashore extends for 850 km, alternating sandy beaches with cliffs. Southern regions are
210 subject to Mediterranean climate, while the north has Atlantic influences (Ferreira, 2000). Mean
211 annual temperature is very mild and ranges from 7.5 °C to 17.5 °C (Ferreira, 2000) and the
212 annual solar radiation reaches one of the highest levels in Europe (Joint Research Centre, n.d.).

213 These topographic and climate characteristics are an invitation to outdoor activities (Santos et
214 al., 2012). As in most Europe and particularly in the last two decades, the demand for outdoor
215 activities and leisure in Portugal shows a growing trend (Roque et al., 2014; Serôdio, 2012), and
216 geocaching is no exception. When this study was conducted circa 35000 geocaches were active
217 in Portugal (“Geopt.org - Portugal Geocaching and Adventure Portal,” n.d.).

218 **2.2. Geocaching data**

219 When finding a geocache, in addition to sign in the logbook hidden in the container, geocachers
220 also can log in the official geocaching website, where anyone can register for free. Many logs
221 simply acknowledge another “found it”, but geocachers are encouraged to share their discovery
222 by posting small texts and photos, building a strong community sense around this recreational
223 activity. Premium members can also vote for favourite caches, leading other geocachers to try
224 to find them afterwards. All geocachers have access to the official website (geocaching.com),
225 where all this information is available.

226 Geocaching data for this study was collected from www.geopt.org (one of the two Portuguese
227 geocaching forums) on November 22nd, 2016 (full dataset). To build a uniform matrix of
228 comparable data, only the traditional geocaches (see Table 1 for terminology) were kept and
229 geocaches with less than 50 logs were excluded.

230 **Table 1 - Geocaching terminology relevant for the present study.**

Term	Description
Geocacher	Person who does geocaching, as opposed to “muggles”, who don’t know the game
Geocache(s)/caches	Hidden container at a specific coordinate ¹ , which has at minimum a logbook for geocachers to sign. There are 18 types of geocaches. Besides the traditional geocache (see definition below), the other types can involve several locations, being the last one the real cache or, for example, a puzzle that the geocacher must solve in order to obtain the container coordinates. ¹ Geocaching has emerged in the year 2000 after the removal of the international degradation of GPS signals (Selective Availability) reducing the error to 10-15 m (http://www.gps.gov).
Owner	Geocacher that creates and places a geocache on the ground for the community.
Traditional cache	This is the original type of geocache and the most straightforward. These geocaches are containers at specific coordinates. The size and contents of the container may vary, but at minimum, all have a logbook. Larger containers may also contain items for trade.

Log	Act of registering a visit to the geocache at the website, even if the geocache was not found. Thus, we used the number of logs as a proxy to the number of visits.
Found/not found	When a geocacher logs a geocache he (she) must register if it was found or not found.
Terrain (TERR)	Physical effort needed to reach a geocache. Provided by the owner of the cache on a scale from 1 (less effort) to 5 (more effort).

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232 **2.3. Land use data**

233 In order to analyse the landscape surrounding each geocache we used a land use map (COS –
 234 *Carta de Uso e Ocupação do Solo* - 2007) of mainland Portugal (DGT – Direção Geral do
 235 Território, 2011) with a minimum mapping unit of 1 ha. This map has a hierarchical classification
 236 from level 1 (minimum detail) to level 5 (maximum detail), where classes of higher detail are
 237 nested within less detailed classes. For this study, 8 major land use classes were considered: i)
 238 Artificial, ii) Agriculture, iii) *Montado*, iv) Open Forests (OF), v) Forests, vi) Shrublands, vii) Open
 239 Areas and viii) Water (Figure 1). Some classes are the same as the defined in level 1 of COS
 240 2007, such as Artificial, while others were defined using higher levels of detail (see table in
 241 Annex for detailed description of land uses). COS 2007 level 3 legend perfectly matches
 242 CORINE Land Cover CLC 2006 level 3 legend, allowing the integration of COS product with the
 243 European CORINE program. The same COS level 3 also matches other international mapping
 244 initiatives like TBFRA 2000 (UN-ECE/FAO, 2000) and LCCS (Di Gregorio et al., 2016). For the
 245 classification of *montado*, all the open forests (canopy cover less than 30%) and agroforests
 246 including cork oaks and/or holm oaks were considered. The category “Open forests” (OF)
 247 corresponds to agroforests including other species and mixes of species different from those
 248 found on the *montado*, such as the pyrenean oak (*Quercus pyrenaica*) and other oaks (*Quercus*
 249 sp.), sweet chestnuts (*Castanea sativa*), eucalyptus or pines. We used this category to compare
 250 the preferences for *montado* landscapes with other structurally similar, assuming that they are
 251 not recognized by Portuguese as having the same cultural value. To extract land use variables,
 252 the land use vector file was converted to a raster file with a 50 m resolution using GRASS (Team,
 253 2015). Although there is a temporal mismatch between landscape data (2007) and geocaching
 254 data (2016), land uses did not change significantly between those years (Caetano et al., 2018).

255 Considering the 8 land use classes defined for this study, it is evident the predominance of
 256 Agriculture (31%) in Portugal, followed by Forests (24%). Shrublands (14%), Open Areas (10%)
 257 and *Montado* (9%) were less represented and some classes (Artificial, OF and Water) covered
 258 less than 5% of the land (Figure 1, Table 3).

259 A 250 m buffer was created around each geocache as a spatial unit to extract land use
 260 geospatial variables: i) land use (LU) at the geocache coordinate; ii) dominant land use in the
 261 buffer (DLU); iii) variety (VAR), i.e., the number of land use classes in each buffer. Average
 262 altitude (ALT) in each buffer was withdrawn from the 90 meter SRTM v4.1 digital terrain model
 263 (Jarvis et al., 2008). To minimise spatial autocorrelation some geocaches were excluded from

264 the analysis. This was done by a hierarchical process in GIS, starting by identifying and retaining
265 the non-overlapping geocaches. Then, a grid with 1km squares was superimposed on the
266 overlapping geocaches and we selected the geocache closer to the centroid in each square.
267 Finally, we merged these to the previously retained non-overlapping caches and attained our
268 final subset. All these analysis were performed using QGIS (QGIS Development Team, 2015).

269 **2.4. Data analysis and statistics**

270 Revealed preferences (following the definition by Cord et al. (2015)) of geocachers for each land
271 use were assessed with Pearson's correlation coefficient by taking the number of caches
272 available on each land use and the number of logs (=visits) to these sites. A positive correlation
273 between the number of available caches and the number of logs would reveal that geocachers
274 simply make more visits to land uses where more caches are available, suggesting that the main
275 reason for the decision of the sites to visit is not related to land use.

276 Stated preferences (again following Cord et al. (2015)) were assessed using a combined
277 approach. In a first step, we considered four variables from the geocaching database expressing
278 distinct attributes of geocacher preferences (Table 2) and those described by absolute values
279 (all except *Log size*) were divided by the number of logs, since the number of logs can be very
280 different among caches. The comparison of these variables, after transformation, between land
281 uses was done using column charts. In a second step, a Principal Component Analysis (PCA)
282 was applied on the four variables and the first axis of the PCA was considered as the new latent
283 variable representing the stated preferences of geocachers. PCA was performed based on the
284 correlation matrix of variables using R version 3.3.2. (R Core Team, 2017). Differences in stated
285 preferences (first axis of the PCA) between land uses were tested with non-parametric tests
286 (Mann-Whitney U tests or Kruskal-Wallis, according to the number of categories), followed by
287 post-hoc tests (with Bonferroni correction), in the case of Kruskal-Wallis. These analyses were
288 performed using the software SPSS (IBM Corp., 2017).

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296 **Table 2 - Variables used to measure stated preferences. a – original range of values, b – range of values after**
 297 **transformation.**

<i>Variable</i>	<i>Description</i>	<i>Range (final dataset)</i>	<i>Transformation</i>
<i>Photos</i>	Average number of photos taken at each geocache and uploaded in the official website.	0 - 1555 a 0 - 5.97 b	Number of photos divided by the number of logs.
<i>Votes</i>	Average number of votes given by geocachers, for each cache, at the GCVote site. GCVote is an extension for the official geocaching site that enables geocachers to rate the quality of caches. Anyone can register for free at this site and vote for any geocache. Rating scale goes from 1 (poor) to 5 (awesome).	0 – 107 a 0 – 0.338b	Number of GC votes divided by the number of logs.
<i>Favourites</i>	Average number of favourites by cache. For every 10 geocaches found, premium geocachers can choose a favourite.	0 – 567 a 0 – 0.683 b	Number of favourites divided by the number of logs.
<i>Log size</i>	Average size of the log (length of the text written during log in, measured by the number of characters) for each cache.	77 – 1256 a	No transformation.

298 We further used a Generalized Additive Model (GAM) (Hastie and Tibshirani, 1990) to
 299 understand which terrain features may influence geocachers stated preferences. GAM has the
 300 advantage over other methods (e.g. Generalized Linear Models), as it allows to model non-linear
 301 relationships between all or some explanatory variables and the response variable. GAM was
 302 estimated by a quadratic penalised likelihood approach, as implemented in the mgvc package
 303 for R (Wood, 2011). Smoothed terms were based on smoothed splines. Four variables
 304 potentially related with geocachers connection with the landscape were used as explanatory
 305 variables: the dominant land use in each buffer (DLU), the number of land uses present in each
 306 buffer (VAR), the mean altitude in each buffer (ALT) and the physical effort needed to reach a
 307 geocache (TERR). These four variables were selected because were suspected to have an
 308 influence on perceptions of the landscape: DLU and VAR influence preferences, ALT influences
 309 the perception of the surroundings and TERR was used to test if a variable non-related to land
 310 use could influence preferences.

311 The first axis of the PCA described in second step was used as the response variable. DLU is
 312 a categorical variable, i.e. coded as dummy variable in the analysis, and hence it was not
 313 included in the GAM as a smoothed term, i.e., only parametric coefficients were computed for
 314 this variable. The *montado* class of the dummy variable was coded as the reference DLU, i.e.
 315 when all other DLUs equals zero, so the resulting coefficients of the parametric terms refers to

316 this class. Because the effect of VAR, ALT and TERR can be potentially influenced by DLU, we
 317 also included the interaction terms between the three variables and DLU in the model.

318

319 3. Results

320 3.1. Revealed preferences

321 3.1.1. Full dataset

322 The full dataset integrated 50818 geocaches, prevailing those located in Artificial (37%),
 323 followed by Forests (19%), Agriculture (17%), Shrublands (12%) and Open Areas (8%). All the
 324 other LU classes had less than 4% of caches each (Table 3). The number of geocaches in each
 325 LU was not proportional to area covered by it, particularly in artificial areas, which have the
 326 highest density of geocaches (4.34 caches/km²), while all the other LUs had less than 1
 327 cache/km².

328 The percentage of geocaches available in each LU type was very similar to the percentage of
 329 logs in the correspondent LU (Table 3), also shown by a very high correlation between the
 330 number of logs and number of geocaches per land use (0.98 Pearson $p < 0.001$), showing that
 331 geocachers tend to visit the caches according to their availability, regardless of the land use.
 332 The differences between these two percentages are less than 3% in every land use class, with
 333 the single exception of Artificial (7.8%), which is the single land use having more visits than
 334 expected. Artificial was the most visited land use (44% of logs) followed by Forest (17% of logs),
 335 and then by Agriculture, Shrublands, Open Areas, Water, OF and *Montado*, in precisely the
 336 same order as for the availability of caches per land use.

337 **Table 3 - Percentage of each land use class on the study area, number and percentage of logs and number, percentage**
 338 **and density (caches/km²) of geocaches from the full dataset in each land use category.**

Land use	Land use Area		Cache density	Geocaches		Logs	
	(%)	(Km ²)	(caches/ Km ²)	N	(%)	N	(%)
Artificial	5	4267	4.34	18525	37	3856168	44
Forests	24	21365	0.45	9616	19	1462276	17
Agriculture	31	27870	0.31	8624	17	1210202	14
Shrublands	14	12181	0.48	5897	12	857746	10
Open Areas	10	9214	0.45	4143	8	704786	8
Water	2	1853	0.84	1556	3	288555	3
OF	5	4425	0.41	1834	4	260024	3
Montado	9	8099	0.08	623	1	81099	1
Total	100	89274	0.57	50818	100	8720856	100

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341 3.1.2. Final subset

342 This dataset included 11335 geocaches, mainly distributed by three LU classes: Artificial (27%),
 343 Forest (22%) and Agriculture (22%). Shrublands had 12% of the geocaches (1377) and only 2%
 344 (207) were in *Montado* (Table 4). Altogether, the remaining LU categories had 15% of the
 345 geocaches. The balanced proportion of geocaches in the three main LU categories results from
 346 the exclusion of spatially correlated geocaches, particularly in artificial areas, where overlap was
 347 more frequent.

348 **Table 4 - Number and percentage of logs and number and percentage of geocaches from the final subset by each land use**
 349 **(LU) and dominant land use (DLU) category.**

	LU (at the geocache)				DLU (for the 250m buffer)			
	Geocaches		Logs		Geocaches		Logs	
Land use	N	(%)	N	%	N	%	N	%
Artificial	3107	27	761205	34	2210	19	625003	28
Agriculture	2524	22	436638	20	3676	32	624411	28
Forests	2506	22	437662	20	2655	23	457870	20
Shrublands	1377	12	248941	11	1256	11	218681	10
Open Areas	961	9	204838	9	676	6	131454	6
OF	419	4	77607	3	276	2	50965	2
Water	234	2	52100	2	393	3	111860	5
Montado	207	2	32339	1	193	2	31086	1
Total	11335	100	2251330	100	11335	100	2251330	100

350 When considering the dominant land use (DLU) around each cache (250 m buffer) the
 351 distribution of caches per class changed slightly (Table 4). Geocaches surrounded mostly by
 352 Agriculture became the most frequent (32%), followed by Forests (23%), Artificial (20%),
 353 Shrublands (11%) and Open Areas (6%). All the remaining DLUs were present in 8% of the
 354 buffers with the *Montado* representing only 2% (193 caches).

355 Geocaches with Agriculture, Forests or Artificial DLU are the most visited, in similar proportions
 356 (20-28%) of the logs, while the remaining land uses are visited according to the availability of
 357 caches. Thus, for our final dataset, although the order of available caches per DLU changed,
 358 the correlation between available geocaches and the number of logs per DLU was still very high
 359 (Pearson 0.94; $p < 0.001$).

360 **3.2. Stated preferences**

361 The four variables considered exposed the preference of geocachers by open and vast land
 362 uses, particularly Open Areas, Water and Shrublands (Figure 2). More photos were taken, and
 363 more favourites were chosen at Water sites while Open Areas were the land use where
 364 geocachers attributed more votes and wrote longer texts. Shrublands were the second DLU
 365 about which geocachers wrote longer texts and the third where they took more photos and voted
 366 more both for Favourites and Votes. Considering forested land uses (Forests, *Montado* and

367 Open Forests) preferences also go for more open habitats, namely for the *Montado*, which was
 368 the second DLU most voted. Finally, a third group of DLUs includes the most human-
 369 transformed landscapes (Artificial and Agriculture), which were always the less preferred for any
 370 of the variables considered.

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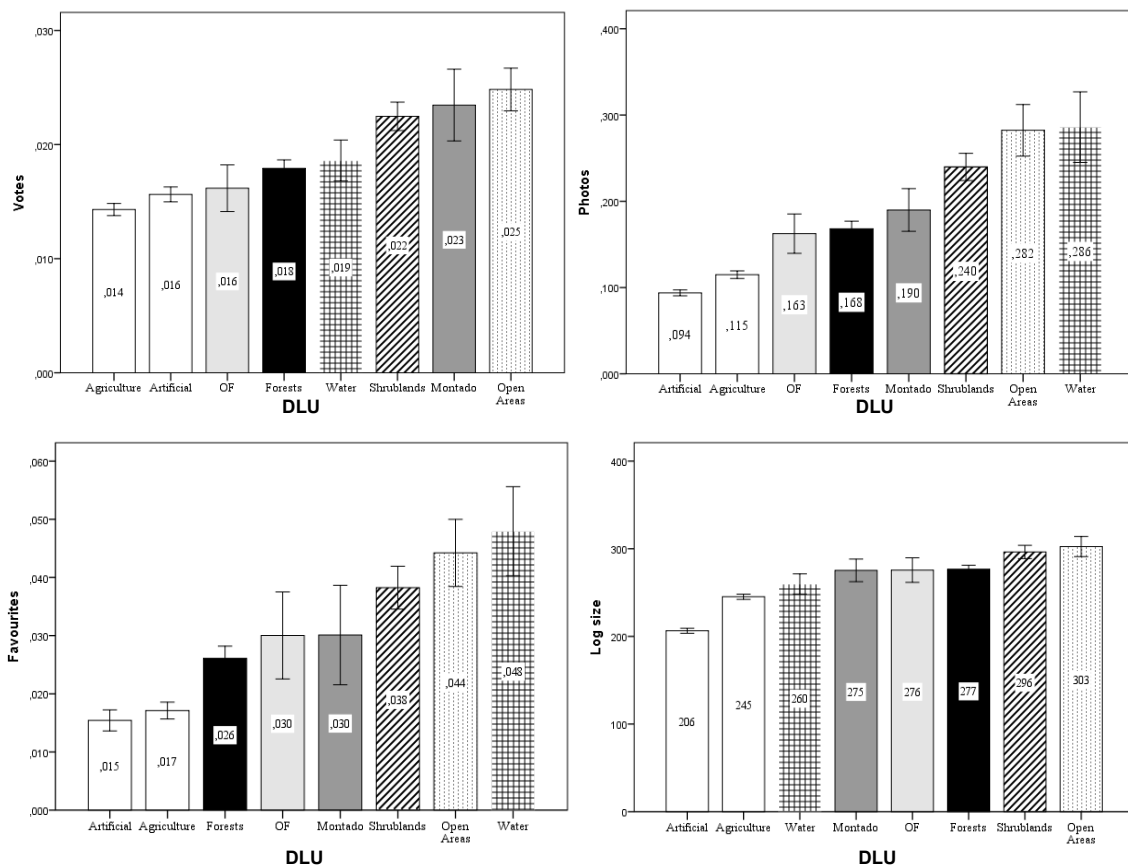
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385 **Figure 2 – Average values (columns) in each DLU for the four variables considered to assess the stated preferences:**
 386 **Photos, Votes, Favourites and log size. Error bars: 95% confidence interval.**

387 The first PCA axis (Comp1) explained 49% of the variance and the second (Comp2) 24% (Table
 388 5).

389 **Table 5 - Factor loadings for Comp1 and Comp2 derived from the principal component analysis (PCA) to show revealed**
 390 **preferences of geocacher for land uses.**

	Comp1	Comp2
Photos	0.584	-0.122
Votes	0.325	-0.861
Favourites	0.557	0.262
Log size	0.492	0.418
Eigenvalues	1.403	0.979
Variance explained (%)	49.2	24.0
Variance accumulated (%)	49.2	73.2

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393 The number of photos, log size and the number of favourites were the variables contributing the
 394 most for Comp1. All variables had positive loadings on the first PCA component. Given this, we
 395 assumed that this component can be interpreted as a stated preferences gradient with higher
 396 values meaning a higher stated preference for a cache.

397 The first two PCA components are shown in Figure 3, where we also identified the three groups
 398 of DLU previously suggested from the less preferred to the most preferred: i) Human-altered -
 399 Artificial and Agriculture land uses, the group with lower stated preferences; ii) Forested -
 400 Forests, OF and *Montado*, with intermediate stated preferences, and finally, iii) Open
 401 landscapes (Water, Shrublands and Open Areas), with the highest stated preferences. The
 402 Forested land use closer to the Open landscapes cluster is the *Montado*.

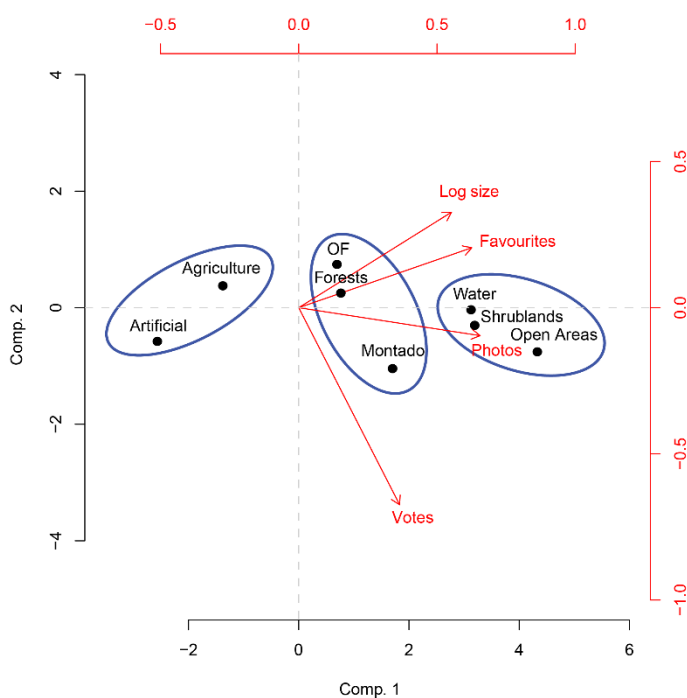


Figure 3 - Biplot of the first two components of PCA (Comp1 and Comp2; explaining 73% of total variance) that represents the factor loadings of stated preferences variables (red labels, arrows and axis) and the mean scores for the categories of Land uses (black labels, dots and axis).

409 3.2.1. Forested land uses (*Montado*, OF, Forests)

410 Within the forested land uses significant differences were found (Kruskal-Wallis test, $p < 0.05$)
 411 when considering the new latent variable representative of the stated preferences, with *Montado*
 412 having the highest Comp1 value (Figure 3). Post hoc tests revealed significant differences
 413 between *Montado* and Forests ($p = 0.001$) and between *Montado* and OF ($p = 0.030$). No
 414 differences were found between Forests and Open Forests.

415 3.2.2. Factors affecting the stated preferences

416 The GAM parametric coefficients reveal that, with the exception of Open Areas, geocachers
 417 tend to prefer areas dominated by *Montado* more than any other DLU class, as shown by the
 418 negative sign of the coefficients (Table 6). These differences are significant for the human-

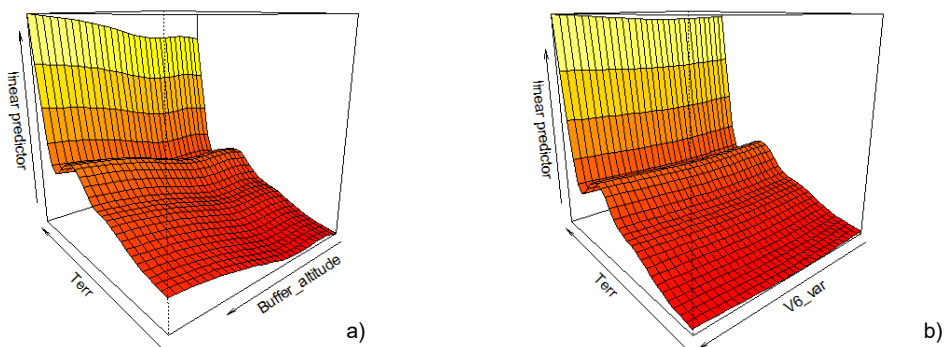
419 altered land uses, Forests and OF, but not for the open landscape uses (Water, Shrublands and
 420 Open Areas).

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422 **Table 6 - Summary of GAM parametric terms (p-value interval codes: *** <0.001, ** < 0.01, * < 0.05, <0.1).**

GAM terms	Estimate	Std. Error	t value	p-value
(Intercept)	0.288	0.092	3.142	0.002 **
DLUArtificial	-0.694	0.099	-7.008	2.56e-12 ***
DLUAgriculture	-0.455	0.094	-4.831	1.38e-06 ***
DLUOpen Areas	0.030	0.105	0.289	0.772
DLUShrublands	-0.095	0.100	-0.948	0.343
DLUOF	-0.242	0.123	-1.969	0.049 *
DLUForests	-0.265	0.095	-2.806	0.005 **
DLUWater	-0.041	0.142	-0.287	0.774

423 Besides DLU, stated preferences of geocachers were also influenced by ALT, TERR and VAR
 424 as revealed by GAM ($R^2 = 0.38$). Overall, geocachers preferred places more difficult to reach
 425 (higher TERR values) and located at higher altitudes (Figure 4a). The influence of TERR is
 426 consistent across the DLUs, as shown by the significance of smoothed terms (Table 7). The
 427 influence of ALT is not as consistent, and it is not statistically significant for OF and water. The
 428 influence of VAR is only statistically significant in areas dominated by open landscape uses
 429 (Water, Shrublands and Open Areas); in these landscapes geocachers prefer less variety of
 430 land uses (Table 7). Nevertheless, this influence is not as clear as for ALT, which means that
 431 the preferences of geocachers are less strongly influenced by variety of land uses (Figure 4b).



433 **Figure 4 – a) and b) 3-dimensional plots representing the joint effect of TERR with ALT and TERR with VAR.**

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437 Table 7 - Summary of GAM smoothed terms and their approximate significance (edf - estimated degrees of freedom;
 438 Ref. df - degrees of freedom for reference distributions; F - F-statistics; p-value interval codes: *** <0.001, ** < 0.01, * <
 439 0.05, . <0.1).

GAM terms	Edf	Ref. df	F	p-value
Interaction TERR x DLU				
s(TERR):DLUMontado	1.997	2.539	11.813	1.22E-06 ***
s(TERR):DLUArtificial	2.506	3.131	15.995	1.44E-10 ***
s(TERR):DLUAgriculture	6.920	6.996	47.123	< 2e-16 ***
s(TERR):DLUOpen Areas	4.107	5.025	189.924	< 2e-16 ***
s(TERR):DLUShrublands	5.244	6.090	181.376	< 2e-16 ***
s(TERR):DLUOF	3.131	3.908	29.057	< 2e-16 ***
s(TERR):DLUForests	6.907	6.995	118.763	< 2e-16 ***
s(TERR):DLUWater	4.079	4.979	121.313	< 2e-16 ***
Interaction VAR x DLU				
s(VAR):DLUMontado	1.000	1.000	0.003	0.954
s(VAR):DLUArtificial	1.000	1.000	2.200	0.138
s(VAR):DLUAgriculture	1.875	2.350	0.896	0.364
s(VAR):DLUOpen Areas	3.572	4.050	8.839	0.000 ***
s(VAR):DLUShrublands	1.060	1.117	8.396	0.003 **
s(VAR):DLUOF	1.584	1.964	0.721	0.522
s(VAR):DLUForests	1.000	1.000	0.625	0.429
s(VAR):DLUWater	4.557	4.890	6.632	0.000 ***
Interaction ALT x DLU				
s(ALT):DLUMontado	1.000	1.000	4.968	0.026 *
s(ALT):DLUArtificial	3.340	3.735	1.718	0.087 .
s(ALT):DLUAgriculture	2.978	3.469	2.273	0.062 .
DLUOpen Areas	4.592	4.912	19.075	< 2e-16 ***
s(ALT):DLUShrublands	1.887	2.363	2.750	0.053 .
s(ALT):DLUOF	2.596	3.110	1.293	0.271
s(ALT):DLUForests	3.298	3.958	7.680	5.12E-06 ***
s(ALT):DLUWater	1.000	1.000	0.023	0.879

440 4. Discussion

441 Our results indicate that overall there is no preference for any land use when geocachers plan
 442 their visit, suggesting that playing the game is their main motivation. However, stated
 443 preferences show that geocachers appreciate more open land uses, exposing the potential of
 444 these crowdsourcing data as CES indicators.

445 4.1. Revealed vs stated preferences

446 Geocache owners hid them in every land use, with the highest density in Artificial areas and in
 447 regions with higher population densities. A relation between cache density and urban areas has
 448 already been shown, although with preference for green spaces (Cord et al., 2015; Santos et
 449 al., 2012). Probably, as in other countries, owners place caches near their residence, although
 450 choosing places with elements that can attract geocachers (Cord et al., 2015).

451 The frequency of visits by geocachers to a site was clearly related to cache availability, which
 452 means that they did not choose *a priori* any LU. Motivations to do geocaching seem to be more

453 related to ludic aspects of the game such as the will to collect or the challenge of being the first
454 one discovering a cache (O'Hara, 2008). Although discovering new places or walking outdoors
455 can also motivate geocachers (O'Hara, 2008), previous studies already showed that geocachers
456 usually do not move far from their own municipality (Cord et al., 2015; Santos et al., 2012).

457 The evidence that geocachers do not show *a priori* preferences for any LU can be viewed as a
458 "natural experiment" when their stated preferences are assessed after the visit. In line with
459 previous studies on landscape appreciation by humans (Hartmann and Apaolaza-Ibáñez, 2010;
460 Kaplan, 1987; Williams and Cary, 2002, 2001), our study showed that in Portugal geocachers
461 prefer the more open land uses and higher places, a preference that may be explained by sense
462 of vastness at the geocache site; when at lower altitudes, and particularly in the case of forested
463 land uses, trees reduce the visual horizon. Geocachers also stated preference for places with
464 water, another common feature of human nature (Petrova et al., 2015; Tieskens et al., 2018)

465 Terrain difficulty was relevant for geocachers; regardless of the land use they showed
466 preference for sites more difficult to reach. The challenge is one important motivation for
467 geocachers, both from the individual and social point of view (O'Hara, 2008). Geocachers that
468 log at caches more difficult to reach attain a personal achievement and at the same time
469 increase their reputation in the community. A general preference for less fragmented areas
470 (fewer land uses within each buffer) although not as strong, has also been identified. More
471 fragmented areas lessen the sensation of vastness and the sense of control of the landscape
472 (de Val et al., 2006), which is in accordance with the preference for more open landscapes.

473 **4.2. The *montado* landscape**

474 Although the *montado* is not the preferred land use when compared to open landscapes,
475 according to stated preferences, when associated with higher altitudes and terrain difficulty
476 becomes as preferred as those open land uses. It is also the preferred among land uses that
477 have trees. Human preference for savannah-like landscapes has been referred in many studies
478 (Falk and Balling, 2010; Hartmann and Apaolaza-Ibáñez, 2010; Williams and Cary, 2002). The
479 innate preference for open landscapes was suggested by the National Research Council (1986)
480 as an evolutionary advantage for hunters and gatherers living on savannahs, at the time when
481 the hominid brain was increasing in size. According to these authors, these landscapes provide
482 the best shelter, hunting and disease-free environments for hominids. The possibility to see
483 potential predators and prey without being seen (Appleton, 1975), or to easily move through the
484 landscape, is also advantageous (Kaplan, 1991). In spite of the appeal of the savannah
485 hypothesis (Dart, 1925), it is also evident that this preference can, and typically is, modified
486 through personal experience and enculturation (Falk and Balling, 2010; Svobodova et al., 2011).
487 People would tend to prefer more familiar land uses as they grow up, and it would be expected
488 that people that live near a particular land use would have a higher preference for that land use

489 as compared to the others. This could not be assessed in our study because geocache
490 databases do not provide personal details of the geocachers. However, the *montado* landscape
491 covers one of the less populated regions of Portugal (only 7,3% of the population lives in Alentejo
492 (PORDATA and INE, n.d.)), and the proportion of geocachers that was born and raised in
493 Alentejo is expectedly small. Most of the Portuguese population lives near the coast, where
494 artificial, agricultural and forest plantations are the dominant land uses, and so it is probable that
495 most of the positive appreciations of *montado* have been stated by geocachers that do not live
496 surrounded by this landscape.

497 The preference for the *montado*, comparable to that for the open landscapes, can be explained
498 by the presence of scattered trees, considered an element of appreciation of landscapes (Cook
499 and Cable, 1995; Kaplan et al., 1989; Kaplan and Kaplan, 1989; Ruddell and Hammitt, W., 1987;
500 Williams and Cary, 2002). However, and very interestingly, the other types of open forests (OF)
501 that may be found in Portugal, which are also open, savannah-like landscapes with scattered
502 trees, are not as appreciated by geocachers as the *montado*. One of the factors that can explain
503 this preference is the spreading shape of the canopies, which is characteristic of evergreen oaks
504 (cork and holm oaks). In effect, the other open forest types have different tree species, such as
505 eucalyptus, pine trees or other conifers with more columnar canopy formats, or broadleaf
506 deciduous trees, such as the sweet chestnut or the Pyrenean oak, which are more rounded in
507 shape. An aesthetic, emotional and physiological preference for spreading canopies, as
508 compared to columnar or rounded canopies, was found previously, with people feeling happier
509 when viewing spreading trees compared to other tree formats (Lohr and Pearson-Mims, 2006),
510 this being consistent with the savannah hypothesis.

511 Another explanation for the preference for *montado* could be the typical presence of free-ranging
512 livestock in this landscape, but we cannot confirm if it occurred by the time geocachers visited
513 the caches. Although the general interest of human beings for life is acknowledged, the aesthetic
514 and emotional value that the animals represent remains rarely discussed (Parsons, 2007).

515 Finally, this preference can be attributed to the branding of this landscape, already considered
516 a national trademark. The name *montado* is the brand for several food products, from sausages
517 to wine, cork products such as wallets or footwear, and even hospitality and tourism, being
518 representative of the *Alentejo* region.

519 **4.3. Geocaching as a crowdsourcing indicator for CES assessment**

520 The value of a landscape to humans is not easy to quantify since many of the services delivered
521 are not products traded on markets. Some services, such as ecotourism or recreation, can only
522 be valued by monetary revealed preferences techniques, such as travel cost methods or
523 hedonic pricing. However, most CES can only be valued by stated preferences methods, where
524 people are asked about their preferences in face of hypothetical changes of the landscape.

525 These economic methods have been criticized since revealed preference based on monetary
526 valuation is dependent on consumers' sovereignty and not on ecological conditions (Kumar and
527 Kumar, 2008; Milcu et al., 2013), and stated preferences monetary valuation (such as
528 Contingent valuation or Choice Experiments) are not real situations and the money people state
529 be willing to pay is not real (Spash, 2007). Given the problems associated with monetary
530 valuation, many authors increasingly focus on non-economic methods (Buchel and
531 Frantzeskaki, 2015; Edwards et al., 2012; Fagerholm et al., 2012; Hermelingmeier and Nicholas,
532 2017; Palomo et al., 2013; Plieninger et al., 2013a, 2013b; Sherren et al., 2010; Williams and
533 Cary, 2002, 2001). In recent years, crowdsourcing indicators are becoming popular (Figueroa-
534 Alfaro and Tang, 2017; Gliozzo et al., 2016; Tenerelli et al., 2016; Yoshimura and Hiura, 2017)
535 given the high number of photographs, the easiness to obtain them and the geographical
536 location they provide. Most studies use the number of photographs taken from a particular site
537 as a proxy for the intensity of the respective CES (Martínez-Pastur et al., 2016; Tenerelli et al.,
538 2016).

539 Geocaching delivers large numbers of photographs available online, and thus can be a good
540 crowdsourcing indicator of CES. As compared to the other user generated contents, geocaching
541 is certainly less known and maybe this is the reason why it is still rarely used as an ES indicator.
542 Nevertheless, the main difference between geocaching and the other user generated contents
543 is that it does not assume that people go to a place attracted by landscape attributes, but rather
544 motivated by the activity itself (which was corroborated in this study). In addition, all participants
545 go to the same specific point where the cache is located rather than being scattered through the
546 landscape. This provides the opportunity to compare information stated by people exactly at the
547 same location, overcoming an often referred problem, which is the error in geotagged photos of
548 other social networks (Tenerelli et al., 2016; Yoshimura and Hiura, 2017).

549 Besides the number of photos taken, other parameters were chosen to evaluate geocachers
550 perceptions about the landscape, such as the number of votes and their scores. These stated
551 preferences have the advantage of using the same scale for all practitioners. However, one of
552 the most promising parameters is the people's writings in the log. Although not explored in this
553 study, the text contents may contain very relevant information about the landscape and about
554 the geocachers' emotional state. Also out of the scope of this study, but with similar information
555 potential and deserving to be further explored, is the description of the cache itself made by the
556 owner and whose information is generally quite extensive and revealing of the choice of location
557 (Mendes et al., 2014).

558 Besides not addressing the information on the text and photo contents, this study has other
559 limitations and it should be considered as a first approach to the use of geocaching to assess
560 CES. To actually understand the feelings of geocachers when at a particular place, further work

561 is needed. For instance, text mining and sentiment analysis, by revealing the emotional polarity
562 of the texts written in the log and the landscape features most times referred (Bermingham et
563 al., 2009; Li and Wu, 2010), would reveal much about connection of geocachers with the
564 landscape. Nevertheless, Fartoukh et al. (2012) found a significant correlation between the
565 number of words and positive emotions in texts written by children with different ages, providing
566 strength to our working hypothesis. Furthermore, the factor loadings of Photos, Log size and
567 Favourites on the first axis of the PCA (Figure 3) are positive and similar, suggesting that long
568 texts are used to describe favourite places. In the same way, as suggested above, the number
569 of photos of a particular landscape is probably positively correlated to a positive emotion but
570 only by analysing the content of the photographs we could understand which are the landscape
571 attributes most captured by geocachers and, consequently their actual preferences (Figueroa-
572 Alfaro and Tang, 2017; Martínez-Pastur et al., 2016).

573 **4.4. Key Insights and Policy Implications**

574 This study reveals that the *montado* is a landscape valued by geocachers, and this may apply
575 to other social groups. *Montado* ecosystem has therefore the potential to provide more
576 recreational and cultural services that what has been acknowledged thus far, since ES for the
577 *montado* remain somewhat poorly studied (Leal et al., 2018)

578 The *montado* faces several threats, some of which are global and common to many ecosystems,
579 such as climate change and increased aridity, with the aggravating factor that almost all the
580 holm (about 99.4%) and cork (about 93%) oak forests are located in areas of high susceptibility
581 to desertification (Dias et al., 2013). Other threats are more specific of the *montado*, such as
582 diseases affecting cork oaks, abandonment or poor management techniques, all of which
583 exacerbated by the dependence of this landscape on the production of cork.

584 Diversification of forestry products and services is one of the strategic objectives for the forest
585 in the National Strategy for Climate Adaptation (Dias et al., 2013), following the Commission
586 *White Paper on Adapting to Climate Change – Towards a European Framework for Action*
587 (2009) and the *EU Strategy on Adaptation to Climate Change* (2013). Several municipalities
588 within the *montado* range already developed strategies for mitigation and adaptation to climate
589 change, including non-structural measures such as information dissemination, economic
590 incentives to reduce vulnerabilities and awareness for adaptation (and against maladaptation),
591 and economic instruments (such as environmental markets).

592 Although these landscapes have a high potential to expand the supply of ES as a response to
593 economic incentives (Bugalho et al. 2017), there is a low uptake of agri-environmental measures
594 for the *montado*-covered areas, mostly due to low compensation values from the landowners
595 perspective (Santos et al., 2015). However, some landowners and managers already regard

596 cultural services as a potential source of income. Their numbers should increase as the
597 recreational or inspirational potential of these landscapes becomes better known.

598

599 **5. Conclusions**

600 In this study, we demonstrated that the information provided online by geocachers can be used
601 to ascertain landscape preferences. Along with expected results, such as the preference for
602 landscapes with water, we have gained insights on the determinants that may explain the
603 attraction for *montado* savannah-like landscapes and this information may support the design
604 of regional development and land-use management policies. Allowing the comparison of *a priori*
605 preferences (choice of place to visit) with the stated preferences (a posteriori site evaluation) is
606 one of the great advantages of using geocaching-produced data, as it allows to assess the
607 effects of any a priori preference for a region or landscape. Other advantage of geocaching data
608 is the reduced error in caches location as compared to other crowdsourcing data and a more
609 numerous and diversified data. On the other hand, when geocachers write their opinions in the
610 logs they are not feeling pressured by anyone, as sometimes happens with face-to-face
611 interviews and other kinds of methods which requires the intervention of a mediator. Our results
612 also indicate that the *montado* has a strong potential as CES provider and, consequently, as a
613 promoter of this landscape diversification, reducing the risks of being highly dependent on
614 provision services. Given the considerable increase of tourism in Portugal, representing 7% of
615 the Portuguese GDP (PORDATA, 2017), this would be of interest and compatible with activities
616 already implemented and, if properly conducted, environmentally sustainable.

617

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628

629

This study	COS description	Agregation
Artificial	Artificial surfaces	level 1
Water	Wetlands	level 1
	Water bodies	level 1
Agriculture	Arable land	level 2
	Permanent crops	level 2
	Pastures	level 2
	Annual crops associated with permanent crops	level 3
	Complex cultivation patterns	level 3
	Land principally occupied by agriculture, with significant areas of natural vegetation	level 3
Montado	Cork or holm oaks agro-forestry with non-irrigated crops	level 5
	Cork or holm oaks agro-forestry with irrigated crops	level 5
	Cork or holm oaks agro-forestry with pastures	level 5
	Cork or holm oaks agro-forestry with permanent crops	level 5
	Cork or holm oak open forests	level 5
	Cork or holm oak open forests with other broadleaf trees	level 5
	Cork or holm oak open forests with evergreen trees	level 5
Forests	Forests	level 2
Open Forests (OF)	Other agro-forests with non-irrigated crops	level 5
	Other agro-forests with irrigated crops	level 5
	Other agro-forests with pastures	level 5
	Other agro-forests with permanent crops	level 5
	Broadleaf open forests	level 5
	Pure or mixed evergreen open forests	level 5
	Evergreen and broadleaf open forests	level 5
Open Areas	Natural grasslands	level 3
	Clear cuts and new plantings	Level 4
	Tree nurseries	Level 4
	Firebreaks	Level 4
	Open spaces with little or no vegetation	Level 2
Water	Wetlands	Level 2
	Water bodies	Level 2

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