

1 **Quantifying wildlife-livestock interactions and their spatio-temporal patterns: is**
2 **regular grid camera trapping a suitable approach?**

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11 **ABSTRACT**

12 Camera trapping use has increased significantly in ecological studies in recent decades
13 due to its ability to register information about cryptic and/or elusive species and, more
14 recently, due to its ability to derive population parameters, such as population abundance
15 or density. For these latter applications, camera traps set in a regular grid pattern (CT-
16 RG) are required to obtain representative information of the study area. The present work
17 aims to assess the usefulness of the information collected through CT-RG to study
18 interspecific interactions between animals, in terms of frequency of interaction and their
19 spatiotemporal pattern. The results from CT-RG were compared with those obtained from
20 GPS collars. For this latter methodology, 31 individuals were monitored with GPS-GSM
21 collars (9 red deer [*Cervus elaphus*], 7 fallow deer [*Dama dama*], 6 wild boar [*Sus scrofa*]
22 and 9 cows). The results showed that all the types of interactions recorded by GPS devices
23 were also recorded by CT-RG. However, the relative frequency of each type of
24 interspecific interaction was not precisely estimated using CT-RG. Nor did we observe

25 an overlap between methodologies in the temporality of the interactions. These results
26 are possibly due to most of the interactions tending to occur at aggregation points, which
27 cannot be sufficiently represented by a regular grid. Finally, the spatial pattern obtained
28 from CT-RG correlated with those obtained with GPS collars. Nowadays, camera
29 trapping is being established as an affordable and effective tool to study different
30 population parameters and, therefore, a huge amount of area is monitored with these
31 devices using regular grids. Our results suggest that the information obtained through CT-
32 RG can also be used to study the patterns of interaction between species.

33

34 KEYWORDS

35 Camera traps; regular grids; GPS; telemetry

36

37 INTRODUCTION

38 The spatial interactions between wildlife species have long been studied to understand
39 different ecological processes, such as animal behaviour, predation and reproduction (Ji
40 et al. 2005). Furthermore, direct and indirect interactions at the wildlife-livestock
41 interface cause conflicts from a socio-economic, ecological and epidemiological point of
42 view (Gortázar et al. 2010). The studies of such interactions in given contexts are useful
43 to understand the actors involved and the potential risk factors explaining the interaction
44 pattern (Perrotton et al. 2017). However, they cannot be generalized to all the interfaces
45 nor to scenarios of the same interface, since spatial heterogeneity drives local differences
46 in community and population processes, which could have consequences in the
47 transmission of pathogens (Real and Biek 2007; Keeling and Rohani 2011). Thus, studies

48 at local scale are necessary to accurately characterize the interaction pattern between
49 species and their spatio-temporal variation.

50 Multiple methodologies have been used to study intra and interspecific interactions
51 between wildlife and domestic animals (reviewed in Triguero-Ocaña et al. in press).
52 Among the most used methods, direct observation, epidemiological questionnaires and
53 individual tracking devices (global positioning system [GPS] and proximity loggers) are
54 the most employed to quantify interactions. Despite their usefulness, these methodologies
55 have some practical limitations, since they are normally expensive, time-consuming,
56 invasive, or limited to large or easily detectable species (Ji et al. 2005; Kukielka et al.
57 2016; Triguero-Ocaña et al. 2019). Therefore, it becomes necessary to test new practical
58 methods and sources of data to characterize the interaction patterns.

59 The use of camera trapping has increased in recent years to monitor wildlife due to its
60 ability to detect elusive species in a non-invasive way (Rovero and Zimmermann 2016).
61 The applications of camera trapping in ecology include, among others, the study of
62 wildlife communities, their activity rhythms and behaviour, population monitoring,
63 occupancy, habitat use and habitat selection (Rovero et al. 2013). This methodology has
64 also been extensively used to determine the abundance and density of species over recent
65 years (Burton et al. 2015). Besides, the recent initiatives of citizen science (i.e. volunteers
66 who participate in the collection of field information for scientific purposes; Cohn 2008)
67 allow an expanding of the spatial coverage of the information collected (Caravaggi et al.
68 2017). Nowadays lots of data are becoming available from camera trapping for large-
69 scale population monitoring, and initiatives at a global scale are emerging to compile the
70 information under quality standards (www.mammalnet.com).

71 Previous studies have already used camera traps (CT hereafter) to record interactions
72 among species (Kukielka et al. 2013; Sparkes et al. 2016). However, these studies focused

73 on the study of interactions at aggregation points or known animal paths in order to
74 maximize the probability of registering contacts. Random sampling design based on the
75 placement of camera traps in regular grids (CT-RG) is often required in many ecological
76 studies (Rowcliffe et al. 2008). This provides the opportunity to explore the potential use
77 of this information (increasingly available and following a grid design) to characterize
78 interactions, particularly at the wildlife-livestock interface, a priority for wildlife and
79 shared disease management (Vicente et al. 2019). In this context, the aim of this work is
80 to assess the potential use of camera traps located in regular grids to study the interspecies
81 interaction pattern.

82

83 MATERIALS AND METHODS

84 The information relating to camera trapping used in the present work proceeds from two
85 studies carried out in Doñana National Park (hereafter, DNP; 37°0' N, 6°30' W; Spain)
86 for the estimation of wild ungulate densities through the utilization of CT-RG. The
87 surveys were conducted in two periods: October – December 2015, and March – April
88 2016. During the 2015, 38 motion infra-triggered camera traps (LTL Acorn, LTL-5310
89 series IR LED Invisible) were deployed (1548 days-camera), and 27 during 2016 (728
90 days-camera). The cameras were set on wooden stakes between 30-50 cm above the
91 ground, without bait, with a separation of 500 m (2015) and 1000 m (2016) and were
92 programmed to record 3 consecutive pictures once the motion sensor was activated.

93 Regarding the reference methodology, during July 2015, 31 ungulates were captured and
94 equipped with GPS-GSM collars in DNP, including 9 adult red deer (*Cervus elaphus*), 7
95 adult fallow deer (*Dama dama*), 6 adult wild boar (*Sus scrofa*) and 9 cows. The captures
96 were performed following the protocol approved by the Animal Experiment Committee

97 of Castilla-La Mancha University and by the Spanish Ethics Committee (PR-2015-03-
98 08). All the collared individuals belonged to different social groups and their home ranges
99 (kernel 95% utilization distribution [UD]) include the area covered by CT-RG surveys,
100 in both periods (see Figure 1). GPS collars were programmed to acquire one geographical
101 location every two hours. The mean positioning error was estimated to be 26 m following
102 the protocol presented by Barasona et al. (2014a).

103 To characterize interactions using camera trapping information, the temporal window
104 between consecutive pictures to be considered as an interaction was defined (Kukielka et
105 al. 2013). It was 2 h, according to the frequency of location fixation of the GPS devices.
106 Using R software 3.5.3 (R Core Team, 2018), for each picture in a specific sampling
107 point, a picture of other species inside the temporal window previously defined was
108 searched and when it appeared, the information of date, time, and the species involved in
109 the interaction were saved. The characterization of interactions through GPS technology
110 was conducted following the methodology presented by Triguero-Ocaña et al. (2019).
111 First, the spatio-temporal window between locations to be considered as an interaction
112 was defined. The spatial window was established as 52 m due to the positioning error of
113 the GPS devices. The temporal window was 2 h according to the frequency of location
114 fixation. Subsequently, using R software, for each relocation the relocation of other
115 individuals inside the spatio-temporal window previously defined was identified and the
116 information of the Euclidean distance between both locations, date and time, plus the
117 coordinates of the interaction were saved.

118 Interactions recorded by both methodologies were compared to assess the capacity of
119 camera trapping to register the interaction pattern, namely the frequency of interactions
120 between pair of species and the spatio-temporal pattern of them in both years. For each
121 pair of species, the frequency recorded by CT-RG was obtained by dividing the number

122 of interactions by the total number of days that all cameras were working. In the case of
123 GPS, the frequency was obtained by dividing the number of interactions by the number
124 of individuals available to interact (i.e. number of collared individuals which had an
125 overlapping home range with the reference individual; see Triguero-Ocaña et al. 2019).
126 Frequencies obtained from GPS refer to available individuals, whereas that obtained from
127 CT-RG refer to the overall population. Therefore, for direct comparisons, frequencies
128 from GPS were corrected for the density of animals in DNP, in that, for each pair of
129 species the frequency was divided by the density of the less abundant species. Densities
130 were 6.3 ind 100 ha⁻¹, 3.9 ind 100 ha⁻¹, 5.7 ind 100 ha⁻¹ and 2.26 ind 100 ha⁻¹, for red deer,
131 fallow deer, wild boar and cattle, respectively (Vicente et al. 2014).

132 In addition, CT-RG has the potential to identify the temporal and spatial interaction
133 patterns. We first calculated the activity pattern of each species with both CT-RG and
134 GPS, then we calculated for each methodology the overlap coefficient between the
135 activity patterns of pairs of species using the R package “overlap” version 0.2.3.
136 (Meredith and Ridout 2014) (see methodological approach in Figure 2a). This overlap in
137 activity patterns can be interpreted as a potential for interaction appearance. To calculate
138 the overlap coefficient between the activity patterns of different species using GPS
139 technology, we followed the protocol presented by Lashley et al. (2018). In a second step
140 we studied the daily variation in the frequency of interactions, and its consistence between
141 methodologies, also using the R package “overlap”. For each year and pair of species, we
142 calculated the coefficient of overlap between the daily patterns in the interaction
143 frequencies described by both methodologies (Figure 2b). Finally, for the spatial pattern
144 we assessed the consistency between methodologies in identifying the most relevant
145 predictors (risk factors) explaining the spatial variation in the interaction frequency. For
146 this purpose, we used the distance to water and the distance to vera ecotone, the most

147 relevant risk factors explaining aggregation of wildlife and livestock in the study area
148 (Barasona et al. 2014b). For each methodology, the frequency of interactions obtained
149 was modelled against the above-mentioned factors using a general lineal model with a
150 binomial distribution and a logit link function (“lme4” R package, Bates et al. 2014). In
151 the case of camera trapping, the response variable was the frequency of interaction (see
152 above) of each camera, and the information of the two environmental variables was
153 calculated, using QGIS 3.4 (QGIS Development Team, 2018), as the average value of
154 each variable contained in a buffer of 1000 m around each camera trap. In the model
155 parameterized with the GPS information we used, as response variable, the number of
156 contacts regarding the number of locations contained in a 100 x 100 m grid covering all
157 the study area and containing the environmental predictors.

158

159 RESULTS AND DISCUSSION

160 Both by CT-RG and GPS technology we were able to record all types of interspecific
161 interactions taking place in our study area. However, we observed significant differences,
162 for both study periods, in the frequency of interactions recorded by CT-RG as regards
163 GPS technology (see Supplementary Material: Figure S1). The differences between
164 methodologies in the frequency of interaction may be due to the lack of representativeness
165 of aggregation points by the CT-RG. The relevance of aggregation points in explaining
166 wildlife abundance and frequency of interactions was previously described in DNP
167 (Barasona et al. 2014b; Triguero-Ocaña et al. 2019). On the contrary, the use of CT-RG
168 would not allow recording of the interactions produced in these areas, but only to record
169 those indirect interactions that occurred due to the movement of different animals on the
170 landscape. In this respect, by reducing the distance between consecutive cameras we
171 could have more opportunities to account for the activity close to aggregation points even

172 when working with a regular grid. Future studies should, therefore, use CT-RG with a
173 shorter distance between devices, or include the simultaneous monitoring of aggregation
174 and random points by means of camera trapping, to test if the discrepancies between rates
175 estimated with CT-CR and those obtained from the reference method (GPS in this case)
176 can be explained by those interactions at the aggregation points. In the absence of this
177 complementary information, the results presented here show the potential of CT-RG to
178 identify the species involved in the interactions process within a multi-host community.

179 Both methodologies registered a high overlap in the activity patterns of all the pairs of
180 species but the obtained overlap indices from CT-RG were not correlated with those from
181 GPS ($\rho = -0.29$, $p = 0.58$). Firstly, this overlap can be interpreted as a measure of the
182 potential that two species have to interact, and therefore, results suggest a high potential
183 for interactions even when CT-RG once more were not able to properly quantify the
184 expected frequency. Surprisingly, these activity patterns were different when comparing
185 methodologies and, in some occasions, the characteristic crepuscular activity peak of wild
186 ungulates was not detected with CT-RG (see Supplementary Material: Figure S2). Similar
187 results were described by Lashley et al. (2018), which observed differences in the activity
188 peaks recorded by camera traps regarding radio-tags. As a step forward, we explored the
189 overlap in the daily variation of the interaction frequency obtained with each
190 methodology and we did not observe a clear overlap. Secondly, the higher frequency of
191 interactions did not correspond to the moments of greatest overlap. These results suggest
192 that the interactions did not occur randomly during the day in response to species activity
193 and probably their frequency is related to the use of key resources where most interactions
194 occur. We also realize that possibly an insufficient sampling effort, mainly in 2016, could
195 have driven the weak relationships detected and therefore we suggest that more studies

196 are needed to disentangle the capability of CT-RG to register the temporal pattern of
197 interactions.

198 Regarding the spatial pattern of interactions, the general linear models showed that there
199 was a greater frequency of interactions around water points and near to vera ecotone (see
200 Supplementary Material: Figure S3 and Table S1). This pattern was not observed for CT-
201 RG in 2016, probably due to the low number of interactions recorded with CT-RG in this
202 period. In fact, by randomly reducing the number of interactions recorded with GPS for
203 this period, we observed that the expected association between the frequency of
204 interaction and the predictors did not always occur (data not included), so we can assume
205 that a greater number of interactions recorded by CT-RG would describe the same spatial
206 pattern and identify the same risk factors, and therefore spatial pattern, as GPS. More
207 studies are needed to verify this claim.

208

209 CONCLUSIONS

210 Camera trapping is currently an affordable and efficient methodology to carry out
211 different studies about population parameters, so it is expected that the information
212 collected through this tool will continue growing over time. The present study aims to
213 assess the potential of CT-RG for studying the pattern of interaction between species.
214 This information can be highly relevant for epidemiological studies carried out in multi-
215 host communities to develop efficient measures for the control of shared diseases. Our
216 first results show that CT-RG can be used to detect the species involved in the interaction
217 process and the spatial pattern of such interactions; however, CT-RG fail in quantifying
218 precise frequencies and describing the temporal pattern. We realize that neither our
219 dataset nor the experimental design was the best to assess the performance of this

220 approach. However, we believe that this note can be a timely first evidence of the potential
221 that this kind of data can have in the future for studying interactions. Further studies in
222 this area are needed to accurately describe the ability of this tool to record intra and
223 interspecific interactions.

224

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301

302 FIGURES

303 Figure 1. Study area, grid of camera traps for the two time periods and home range (kernel
304 95%) of targeted individuals collared with GPS technology.

305 Figure 2. Methodological approach to study the agreement between methodologies
306 recording the temporal pattern of interactions. A) Overlap between the activity pattern of
307 different species and the expected number of interactions (boxplot) grouped in four time
308 periods (Hour1: 0:00 – 5:59, Hour2: 6:00 – 11:59, Hour3: 12:00 – 17:59, Hour4: 18:00 –

309 23:59); this approach has to be calculated for both methodologies. B) Overlap plot of the
310 daily interaction frequency recorded by methodology.

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