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1 **A holistic model to assess risk factors of fasciolosis in** 2 **Ankole cattle**

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14

15 **Abstract**

16 In recent decades, remote sensing (RS) technology and geographical information systems
17 (GIS) were increasingly used as tools for epidemiological studies and the control of zoonotic
18 diseases. Fasciolosis, a zoonotic disease caused by a trematode parasite (*Fasciola* spp.), is a
19 good candidate for the application of RS and GIS in epidemiology because it is strongly
20 influenced by the environment, i.e. the habitat of the intermediate host. In this study, we
21 examined variables which may increase the fasciolosis risk of Ankole cattle in the degraded
22 and overgrazed Mutara rangelands of north-eastern Rwanda. The risk variables considered
23 included three environmental variables (normalized difference vegetation index, NDVI;
24 normalized difference moisture index, NDMI; normalized difference water index, NDWI),
25 two landscape metric variables (rangeland proportion, building density), two geological
26 variables (poorly-drained soil proportion, elevation) and three animal husbandry variables
27 (herb size, adult proportion and the body condition score). *Fasciola* spp. prevalence was used
28 as the dependent variable, sampling season as a fixed factor and four principal components
29 (PCs, condensed from the ten risk variables) as covariates in a univariate General Linear
30 Model. *Fasciola* spp. prevalence was positively correlated to rangeland proportion, cattle

31 herd size in rural areas, adult proportion and individual body condition. Moreover, high
32 *Fasciola* spp. prevalence was found in densely vegetated areas with high moisture (high
33 values of NDVI and NDMI), in combination with large proportions of poorly-drained soil at
34 low elevations. Future investigations should focus on increased sampling across the Mutara
35 rangelands to prepare a predictive, spatial fasciolosis risk map that would help to further
36 improve sustainable land-use management.

37

38 **Key words:** *Fasciola*, Geographic Information System, Remote sensing, risk model, cattle
39 husbandry, environmental factors, Rwanda

40 **1. Introduction**

41 During the past few decades, multi-disciplinary approaches to carry out epidemiology studies
42 using remote sensing (RS) and geographical information systems (GIS) have been
43 extensively applied (Hay, 2000; Thomson and Conner, 2000; Hendrickx et al., 2004; Kitron
44 et al., 2006). Advances in RS have provided the ability to obtain a variety of environmental
45 parameters (e.g. normalized difference vegetation index, NDVI; normalized difference
46 moisture index, NDMI; normalized difference water index, NDWI) with numerous spatial
47 and temporal resolutions that can be related to disease outbreaks and vector distribution (Hay
48 et al., 1997; Robinson, 2000). GIS allows computer-based analysis of multiple layers of
49 digital mapped data, such as satellite sensor data, maps of host populations, vector and
50 disease distributions (Malone et al., 1997; Malone and Yilma, 1999). For example,
51 environmental RS indices and GIS technologies have been applied to identify habitats of
52 parasites and their vectors, such as mosquito-borne diseases (e.g. malaria, Rift Valley fever
53 and dengue); snail-borne diseases (e.g. schistosomiasis and fasciolosis); or tick-borne
54 diseases (e.g. boreliosis; Hay, 2000; Omumbo et al., 2002; Hendrickx et al., 2004; Tatem et
55 al., 2004; Charlier et al., 2014).

56

57 Bovine fasciolosis is a zoonotic disease affecting the liver of wild and domestic ruminants,
58 caused by parasitic trematodes of the genus *Fasciola* (*F. hepatica* or *F. gigantica*). The adult
59 parasite lives in the bile ducts of the hosts' liver and causes substantial financial losses to
60 pastoralist communities worldwide by negatively affecting growth rates and productive
61 parameters (McCann et al., 2010; Byrne et al., 2016). In the early 2000s, global economic
62 losses due to fasciolosis exceeded US\$200 million, with about 300 million cattle infected
63 (Mas-Coma et al., 2005; Dutra et al., 2010). Moreover, owing to the fact that *Fasciola* spp.
64 also infect humans (currently 2.4 to 17 million people are infected with *F. hepatica*; Eslami et

65 al., 2009), fasciolosis represents a serious public health threat, especially in developing
66 countries (Rokni et al., 2002; Mas-Coma et al., 2009). The parasite occurs primarily in
67 swampy areas or on flooded pastures, i.e. the preferred habitat of the intermediate host
68 (pulmonate freshwater gastropods of the family Lymnaeidae; Brown, 2005; Torgerson and
69 Claxton, 1999). Bovine livestock usually become infected by eating water plants and grass
70 from inundated lawns, or simply by drinking contaminated water (Witenberg, 1964). Since
71 the parasite is strongly influenced by the environment, i.e. the habitat of the intermediate host
72 and by the relative longevity of the parasite inside the mammalian host, fasciolosis is an ideal
73 candidate for the application of RS and GIS (Malone and Yilma, 1999).

74

75 Thus, the purpose of this study was to generate a risk model to better understand the
76 epidemiology of *Fasciola* spp. in the degraded and overgrazed Mutara rangelands of north-
77 eastern Rwanda. Hereby, we considered it imperative to follow a holistic approach and to
78 relate parasite data to three environmental variables (NDVI, NDMI and NDWI), two
79 landscape metric variables (rangeland proportion and building density), two geological
80 variables (proportion of poorly-drained soil and elevation) and three animal husbandry
81 variables (herd size, adult proportion and body condition score, BCS) to identify the major
82 risk factors of *Fasciola* spp. infection in Ankole cattle.

83

84 **2. Material and Methods**

85 *2.1 Study area*

86 The Mutara rangelands are located in the Nyagatare District in north-eastern Rwanda (Fig. 1).
87 They are characterised by a tropical rainfall pattern (wet season: March to May and October
88 to November) with an average annual precipitation of 827 mm and a mean annual ambient
89 temperature of 26.5°C. The Mutara rangelands comprise vast open grasslands, interspersed

90 by evergreen bushland and thicket (Kindt et al., 2014) and are traditionally used to graze
91 cattle. Today, the Mutara rangelands harbour an estimated 160,000 cattle, resulting in a cattle
92 density of 81 individuals/km² (Wronski et al., 2017). Moreover, in a significant part of the
93 Mutara, increasing subsistence agriculture and urbanization, leaving only 13% of the total
94 land area in a natural state (CIRAD, 2002; Wronski et al., 2017).

95

96 *2.2 Study animals and faecal sampling*

97 Faecal samples were taken from Ankole cattle, a breed derived from the Sanga type cattle
98 predominantly found in East-Central Africa (Epstein, 1957). Faecal samples were collected at
99 the end of the short and long wet season, i.e. from 19 February to 17 March and from 12 June
100 to 17 July 2016, respectively. In total, 570 faecal samples were obtained from 142 cattle
101 herds. Sampled individuals were randomly encountered along three 2.5 km wide transect
102 belts (22.5, 32.5 and 37.5 km long) stretching between the Tanzanian border (or the border of
103 the modern Akagera NP) in the East and the Ugandan border (or the Byumba Escarpment) in
104 the West (Fig. 1). Three to five faecal samples were collected from each herd directly after a
105 focal animal had defecated. Additionally, coordinates were recorded using an Etrex 20x GPS
106 (Garmin, USA). Faecal samples (30 g/individual) were retained in labelled plastic containers
107 and preserved in 5-10% formalin prior to processing in the laboratory. Since most herds
108 comprised of only females and their offspring (bulls are usually kept in the kraal), only
109 females and their calves were sampled. Exotic Friesians (or hybrids with Ankole cattle),
110 individuals treated with flukicides during the last six months prior to sampling, or individuals
111 that did not experience similar husbandry conditions (e.g. overnight kraaling) were also
112 excluded from our sampling. Moreover, date, time, age composition (number of adults and
113 juveniles), herd size and the BCS of each sampled individual were recorded.

114

115 *2.3 Coprological examination*

116 Faecal samples were processed in the Veterinary Laboratory of Nyagatare Campus,
117 University of Rwanda. A modified sedimentation technique was employed to detect the eggs
118 of *Fasciola* species. In brief, faecal samples (app. 10 g) were crushed, diluted with 140 ml of
119 saturated NaCl solution and filtered. The faecal suspension was transferred into a 15 ml test
120 tube and sedimented for 20 minutes. Subsequently, the supernatant was discarded and the
121 sediment was conveyed to a microscope slide using a pipette. To ease *Fasciola* spp. egg
122 identification, a drop of Methylene blue was added and eggs were counted using a compound
123 microscope with a 10× and 40× magnification (Hansen and Perry, 1994; Mwabonimana et al.,
124 2009; Rojo-Vázquez et al., 2012). The identification of trematode eggs was facilitated by
125 identification keys provided in Hansen and Perry (1994). *Fasciola* spp. prevalence was
126 established as the number of infected individuals divided by the total number of samples
127 taken in each herd (Margolis et al., 1982).

128

129 *2.4 Image acquisition and processing*

130 Four multispectral Sentinel-2 satellite images (European Space Agency, ESA) covering the
131 entire Mutara rangelands (WGS 84, UTM zone 35S, EPSG code: 32735) during the infection
132 season (i.e. the last wet season prior to faecal sampling) were downloaded from USGS Earth
133 Explorer (<https://earthexplorer.usgs.gov>, last accessed on February 2019). Two images were
134 taken during the short wet season (25 November 2015 and 24 January 2016), another two
135 during the long wet season (14 March and 23 April 2016). All images had good quality, i.e.
136 with little or no cloud cover, and were processed using QGIS (version 2.8.6). Prior to index
137 calculation all sampling points were buffered using a radius of 1km as a proxy for the
138 potential activity range of sampled herds (based on the average distance to the next water
139 source; Apio, pers. comm.). Within these buffers, the NDVI, NDMI and NDWI (based on 10

140 × 10 m pixel size spatial resolution at earth surface) were established and averaged to obtain
141 one value for each cattle herd.

142

143 *2.5 Landscape metric variables*

144 The rangeland proportion within each buffer area was determined using a high resolution
145 Google satellite image. Along four radii (i.e. in direction to North, East, South, and West) in
146 each buffer area, the distances intercepted by rangeland or agricultural fields were
147 established. Subsequently, the rangeland proportion for each buffer was calculated. Building
148 density was based on building registration data downloaded from ‘Geofabrik’ open street
149 map (<http://download.geofabrik.de>, last accessed on February 2019). The number of
150 buildings was established for each buffered area and divided by the total buffer size to obtain
151 building density. Large scale rangeland grazing was reported to increase the *Fasciola* spp.
152 infection risk by cattle being more exposed to vegetation contaminated by metacercaria
153 (Kanyari et al., 2010; Murray and Daszak, 2013). By contrast, in more urbanized areas,
154 livestock is mainly fed on freshly cut grass or agricultural waste and therefore exposed to a
155 lower risk of *Fasciola* spp. infection (Kanyari et al., 2010).

156

157 *2.6 Geological variables*

158 Soil data were extracted from the pedological map of Rwanda (Van Ranst and Delvaux,
159 2000). The map was georeferenced and reclassified into two classes, i.e. poorly-drained soils
160 *versus* well-drained soils. Buffered sampling areas were subtracted from the reclassified soil
161 map and the proportion of poorly-drained soil was calculated. Elevation data were collected
162 at each faecal sampling point using a hand-hold GPS. Alluvial soils and low elevations are
163 usually associated to poor drainage and extended periods of flooding, prevalent in areas that

164 correspond to increased *Fasciola* spp. prevalence in cattle (Zukowski et al., 1991; Malone
165 and Yilma, 1999; McCann et al., 2010; Dutra et al., 2010; Bennema et al., 2011).

166

167 *2.7 Animal husbandry variables*

168 For each sampled cattle herd, the herd size and the adult proportion (i.e. the number of adults
169 older than 24 months, divided by total herd size) were established. A visual BCS assessment
170 based on estimating the presence or absence of musculature and fat deposition on the spinal
171 and caudal vertebrae (El Alqamy, 2013) was applied to each sampled individual and
172 subsequently a herd BCS was calculated by averaging scores. Livestock with deprived health
173 condition or a poor nutritional status show usually a poor BCS, thus expecting a high
174 *Fasciola* spp. prevalence.

175

176 *2.8 Data analysis*

177 Absolute data, were log-transformed, whereas relative data were arcsine square root
178 transformed. To standardize data dimensionality, z-score normalisation was applied to the
179 overall data set. The ten independent variables were reduced using Principal Component
180 Analysis (PCA) resulting into four principle components with an Eigenvalue > 1.0,
181 demonstrating 82.21% of the total variance. A univariate General Linear Model (GLM) was
182 used to examine the impact of these risk factors on the *Fasciola* spp. prevalence in Ankole
183 cattle by using the four PCs as covariates. Initially, all two-way interaction effects of all PCs,
184 as well as a fixed factor (sampling season) and a random factor (herd ID) were included into
185 the GLM, followed by a step wise backwards elimination procedure ($p > 0.1$) to omit non-
186 significant interaction effects (all excluded interactions: $F < 1.50$, $p > 0.22$). Effect strengths
187 were established as Wilk's partial eta-squared (η^2). All statistical analyses were carried out
188 using *RStudio* (version 3.5.1)

189

190 **3. Results**

191 In total, 569 individuals from 142 cattle herds were sampled. Out of these, 113 individuals
192 from 70 herds were detected positive for fasciolosis, corresponding to a total animal
193 prevalence of 19.9% and a herd prevalence of 49.3%. Factor reduction using PCA of ten
194 independent variables yielded four Principal Components (PCs, Table 1). PC1 obtained high
195 factor loadings from NDVI, NDMI and NDWI, suggesting that areas covered by dense,
196 woody vegetation corresponded to a high content of moisture in vegetation and soil and to
197 only a few open water bodies. PC2 received high factor loadings from rangeland proportion,
198 cattle herd size and building density, suggesting decreasing cattle herd size in areas where the
199 original savannah vegetation was transformed into fields and human settlements. PC3
200 received high factor loadings from elevation and the proportion of poorly drained soil,
201 indicating that poorly drained soils predominantly occur in areas of low elevation. PC4
202 obtained high factor loadings from the BCS and the proportion of adult individuals in the
203 herd, suggesting that adult animals have generally a better body condition than juveniles.

204

205 A univariate GLM revealed that *Fasciola* spp. prevalence was significantly affected by
206 several independent variables with a main positive effect of PC2, a main negative effect of
207 PC4 (Table 2, Fig. 2a, b), and the interaction effect of 'PC1×PC3' and 'PC2×PC4' (Table 2,
208 Fig. 3a, b). The *Fasciola* spp. prevalence showed no difference between the two sampling
209 seasons (Table 2). Plotting the interaction effect 'PC1×PC3' generated two different slopes
210 when dividing the data by the median of PC1. In the cohort of data with values loading on
211 PC1 larger than the median (i.e., comparatively high NDVI, high NDMI, but low NDWI), we
212 found *Fasciola* spp. prevalence to decrease with increasing values of PC3 (high elevation and
213 well-drained soil, $R^2 = 0.023$; Fig. 3a). However, in the cohort of data with values of PC1

214 smaller than the median (i.e., comparatively low NDVI, low NDMI, but high NDWI) no such
215 effect was found between *Fasciola* spp. prevalence and increasing PC3 ($R^2 < 0.001$; Fig. 3a).

216

217 Plotting the interaction effect 'PC2×PC4' also generated two different regressions when the
218 data were separated by the median of PC2. In the cohort of data with values loading on PC2
219 larger than the median (i.e., comparatively high proportion of rangeland, large cattle herd size
220 and low building density), *Fasciola* spp. prevalence slightly increased with increasing values
221 of PC4 (adult proportion and averaged BCS, $R^2 = 0.006$; Fig. 3b), while in the cohort of data
222 with values loading on PC2 smaller than the median (comparatively low proportion of
223 rangeland, small cattle herd size and high building density), a strong effect was revealed
224 between *Fasciola* spp. prevalence and the increasing PC4 ($R^2 = 0.249$; Fig. 3b).

225

226 **4. Discussion**

227 The overall *Fasciola* spp. prevalence observed in our study (49.3%) was relatively high
228 compared to other studies on cattle in Rwanda (40.2%; Habarugira et al., 2016) or Ethiopia
229 (32.3%; Bekele et al., 2010). However, depending on regional and seasonal factors the
230 prevalence can vary considerably (Habarugira et al., 2016). Moreover, the *Fasciola* spp.
231 prevalence was positively affected by PC2 (landscape metric variables and herd size, Table 2,
232 Fig. 2a), indicating that large areas of original savannah vegetation, overgrazed by large cattle
233 herds facilitated the spread of *Fasciola* spp.. Large stocking rates were previously reported to
234 be the main reason for increased *Fasciola* spp. prevalence in cattle (Howell et al., 2015) and
235 Morgan et al. (2006) suggested that low stocking rates are the prime measure to control the
236 parasite in open grassland. Moreover, these grasslands are located in remote, rural areas with
237 low building density, while in more urbanised areas with higher building densities and more
238 agriculture, cattle herds are smaller and *Fasciola* spp. prevalence tends to be lower. The

239 second main effect on *Fasciola* spp. prevalence was PC4 (adult proportion and BCS, Table 2,
240 Fig. 2b), suggesting *Fasciola* spp. prevalence to decrease with increasing adult proportion
241 and a high BCS. Moreover, the interaction effect of ‘PC2×PC4’ on the *Fasciola* spp.
242 prevalence further highlighted how the landscape metric variables interacted with animal
243 husbandry variables (Table 2, Fig. 3b). In urbanised, agricultural areas with a smaller
244 proportion of rangeland and smaller cattle herds, but higher number of buildings (PC2 <
245 median), the *Fasciola* spp. prevalence decreased with increasing adult proportion and higher
246 BCS (Fig. 3b). However, there was no relationship between *Fasciola* spp. prevalence and
247 PC4 (adult proportion and BCS) in rural areas with comparatively larger proportion of
248 rangeland, larger cattle herds and lower building densities (PC2 > median; Fig. 3b). Here,
249 cattle were heavily infected with *Fasciola* spp., regardless of age and body conditions. This
250 result suggested that the animal husbandry variables (adult proportion and BCS) negatively
251 correlated to *Fasciola* spp. prevalence only in urban areas, where cattle was fed on freshly cut
252 grass or agricultural waste and thus interrupting the parasites’ life cycle. Land use changes in
253 recent years led to increased urbanisation and the transformation of natural savannah
254 vegetation into agricultural land (CIRAD, 2002; Wronski et al., 2017), reducing the
255 availability of grassland for pastoralists and their cattle and thus reinforcing the negative
256 effects of overstocking and overgrazing (Pandey et al., 1993; Taj et al., 2014).

257

258 Our GLM further revealed an interaction effect of ‘PC1×PC3’ on the *Fasciola* spp.
259 prevalence (Table 2, Fig. 3a). Here, *Fasciola* spp. prevalence was not influenced by the
260 geological variables if recorded in areas with comparatively less vegetation, less moisture but
261 more open water bodies (PC1 < median, Fig. 3a). However, in dense woody vegetated areas
262 with high moisture and few open water bodies (PC1 > median, Fig. 3a), high *Fasciola* spp.
263 prevalence was correlated to poorly-drained soil and low elevation. This finding corresponds

264 to our prediction that *Fasciola* spp. was prevalent in well-vegetated areas with high soil
265 moisture and large proportions of poorly-drained soils at low elevations (Tum et al., 2004).
266 Such specific environmental factors of the micro climate affect the presence and abundance
267 of the intermediate host of *Fasciola* spp. (snails of the family Lymnaeidae) and thus
268 determine the life-cycle of the parasite (Mzembe and Chaudhry, 1979; McCann et al., 2010;
269 Charlier et al., 2014).

270 Given results from previous studies (Yilma and Malone, 1998; Malone et al., 1998; McCann
271 et al., 2010; Kantzoura et al., 2011; Portugaliza et al., 2019), areas with sufficient vegetation
272 (high NDVI), high moisture (high NDMI) or numerous open water bodies (high NDWI), i.e.
273 areas facilitating the development of eggs, the mobility of miracidiae and the spread of
274 cercariae, would be expected to show increased *Fasciola* spp. prevalence. Such areas include,
275 flood plains and riverine forest, but also human-modified landscape elements like dams,
276 swamps, ponds and irrigation canals. Dense vegetation with high soil moisture is known to be
277 the ideal snail habitat (Tum et al., 2004; Malone, 2005), and remote sensing indices, such as
278 NDVI and NDMI, were frequently used to assess the environmental variables typical for snail
279 habitats, to identify high risk *Fasciola* spp. areas and to develop regional fasciolosis risk
280 maps (Malone et al., 1998; Durr et al., 2005). However, the hypothesis that high *Fasciola*
281 spp. prevalence occurs in areas with a high density of open water bodies, i.e. a high NDWI,
282 was not proven by our study.

283

284

285

286 **5. Conclusions**

287 The prevalence of *Fasciola* spp. in Ankole cattle was, at least to a certain degree, defined by
288 all independent variables included in our study. In contrast to other gastro-intestinal parasites
289 (e.g. *Eimeria* spp. or strongyle-type nematodes), the intermediate host and the free-living

290 stages of *Fasciola* spp. require habitats covered by dense and lush vegetation with large
291 proportions of poorly-drained soils at low elevations. Therefore, such habitats should be
292 considered as high fasciolosis risk areas for grazing cattle. Our results further confirmed that
293 the land use changes of the Mutara rangelands in recent decades, i.e. increased urbanization
294 and subsistence agriculture, correspond to a reduced availability of space for the pastoralist
295 community, leading to increased overstocking and overgrazing and thus making the Mutara
296 rangelands an unbalanced and unhealthy ecosystem (e.g. increased fasciolosis). In the future,
297 more random sampling across the Mutara rangelands (or the entire country) is needed to
298 prepare a predictive, spatial fasciolosis risk map, which would help to monitor *Fasciola* spp.
299 dispersal routes, and to develop sustainable land-use management strategies that improve the
300 health of humans, their livestock and the ecosystem in which they live.

301

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311

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453

454 **Figure legends**

455

456 **Fig. 1** Location of the study area (three 2.5 km wide transect belts) in Nyagatare District in northern
457 Rwanda. Each sampling location (dots) was aligned to a buffer area of 1km radius (upper right inset)
458 for which independent variables were determined.

459

460 **Fig. 2** The relationships of PC2 (a) and PC4 (b) with the *Fasciola* spp. prevalence in Ankole cattle on
461 the Mutara rangelands.

462

463 **Fig. 3** a. Scatter plot showing the interaction effect of “PC1×PC3” on *Fasciola* spp. prevalence: No
464 relation was unrevealed between *Fasciola* spp. prevalence and increasing PC3 as seen in case of the
465 data with values loading on PC1 smaller than the median (shaded dots, grey dashed line; linear
466 regression: $R^2 < 0.001$), while decreasing *Fasciola* spp. prevalence with increasing values of PC3
467 become evident for the data with values loading on PC1 larger than the median (bold dots, black line;
468 linear regression: $R^2 = 0.023$).

469 b. Scatter plot showing the interaction effect of “PC2×PC4” on *Fasciola* spp. prevalence: Distinctly
470 decreasing *Fasciola* spp. prevalence with increasing PC4 is seen in case of the data with values
471 loading on PC2 smaller than the median (shaded dots, grey dashed line; linear regression: $R^2 = 0.249$),
472 while slightly increasing *Fasciola* spp. prevalence with increasing values of PC4 become evident for
473 the data with values loading on PC2 larger than the median (bold dots, black line; linear regression:
474 $R^2 = 0.006$).

475

476

477

478 **Tables**

479

480 **Table 1** Axis loadings of four principal components (demonstrating 82.21 % of the total variance),
 481 obtained from principal component analysis of ten independent variables (see section 3.2). PC
 482 loadings > |0.5| are shown in bold font type.

483

Principal component	PC1	PC2	PC3	PC4
Eigenvalue	3.21	2.20	1.46	1.36
Percent variance	26.290	25.62	16.15	14.15
NDVI	0.953	0.219	-0.066	-0.032
NDMI	0.894	-0.163	-0.240	0.054
NDWI	-0.892	-0.275	-0.092	0.039
Rangeland proportion	0.170	0.934	0.060	-0.031
Herd size	-0.042	0.872	0.124	-0.100
Building density	-0.187	-0.863	0.229	0.012
Elevation	0.064	-0.117	0.873	-0.041
Poorly drained soil proportion	0.230	-0.117	-0.840	-0.022
BCS	-0.047	-0.019	0.039	0.839
Adult proportion	0.036	-0.083	-0.060	0.832

484

485 **Table 2** Results of the univariate GLM using the *Fasciola* spp. prevalence as the dependent variable,
 486 sampling season as a fixed factor and the four principal components (PCs) as covariates. Insignificant
 487 interaction effects were excluded if $p > 0.1$.

488

variables	Estimate	SE	<i>t</i>	<i>p</i>	Partial eta ²
Sampling season	-0.286	0.170	-1.680	0.095	0.021
PC1	-0.058	0.085	-0.687	0.493	0.008
PC2	0.248	0.083	2.967	0.004	0.080
PC3	-0.065	0.076	-0.850	0.397	0.005
PC4	-0.215	0.081	-2.654	0.009	0.095
PC1×PC3	-0.184	0.083	-2.232	0.027	0.036
PC2×PC4	0.245	0.072	3.383	<0.001	0.079

489

490