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Using road patrol data to identify factors associated with carnivore roadkill counts

Samual T. Williams^{1,2,3}, Wendy Collinson⁴, Claire Patterson-Abrolat⁴, David G. Marneweck^{4,5}, Lourens H. Swanepoel¹

¹ Department of Zoology, School of Mathematical & Natural Sciences, University of Venda, Thohoyandou, South Africa

² Department of Anthropology, Durham University, Durham, United Kingdom

³ Institute for Globally Distributed Open Research and Education (IGDORE)

⁴ Endangered Wildlife Trust, Ardeer Road, Johannesburg 1645, South Africa

⁵ Centre for Wildlife Management, University of Pretoria, South Street, Pretoria 0186, South Africa

Corresponding authors: Samual T. Williams; Lourens H. Swanepoel

Email addresses: samual.t.williams@gmail.com; lourens.swanepoel.univen@gmail.com

18 Abstract.

19 As the global road network expands, roads pose an emerging threat to wildlife populations. One
20 way in which roads can affect wildlife is wildlife-vehicle collisions, which can be a significant
21 cause of mortality through roadkill. In order to successfully mitigate these problems, it is vital to
22 understand the factors that can explain the distribution of roadkill. Collecting the data required to
23 enable this can be expensive and time consuming, but there is significant potential in partnering
24 with organisations that conduct existing road patrols to obtain the necessary data. We assessed
25 the feasibility of using roadkill data collected daily between 2014 and 2017 by road patrol staff
26 from a private road agency on a 410 km length of the N3 road in South Africa. We modelled the
27 relationship between a set of environmental and anthropogenic variables on the number of
28 roadkill carcasses, using serval (*Leptailurus serval*) as a model species. We recorded 5.24 serval
29 roadkill carcasses/100 km/year. The number of carcasses was related to season, the amount of
30 wetland, and the number of owls killed on the roads, but was not related to any of the
31 anthropogenic variables we included. This suggests that roadkill patterns may differ greatly
32 depending on the ecology of species of interest, but targeting mitigation measures where roads
33 pass through wetlands may help to reduce serval roadkill. Partnering with road agencies for data
34 collection offers powerful opportunities to identify factors related to roadkill distribution and
35 reduce the threats posed by roads to wildlife.

36

37 1. Introduction.

38 Roads are integral to a country's continued development and prosperity, especially developing
39 countries such as South Africa (Karani, 2008; Keshkamat, Looijen & Zuidgeest, 2009).
40 However, traffic can also have a direct negative impact on both people and wildlife (Verster &
41 Fourie, 2018), with many species at risk from wildlife-vehicle collisions (WVCs), often resulting
42 in an animal's death, or 'roadkill'. For the last three decades the field of road ecology has
43 highlighted the negative impacts that roads and their associated users have on biodiversity, and
44 their potential to limit wildlife populations (Kioko et al., 2015; Laurance et al., 2015;
45 Parchizadeh et al., in press).

46

47 Infrastructure development impacts biodiversity and ecosystems through exposing ecological
48 habitats to disturbance and fragmentation (Benítez-López, Alkemade & Verweij, 2010), land-use
49 and land cover change, and limiting connectivity in the landscape (Perz et al., 2013; Liang et al.,
50 2014). Effective mitigation efforts required to reduce such impacts often require a detailed
51 understanding of the factors driving wildlife-vehicle-collisions (Gunson & Teixeira, 2015). In
52 developed countries such as the USA and Europe there have been several studies that
53 successfully quantified the factors affecting wildlife-vehicle-collisions (Gunson & Teixeira,
54 2015; Rytwinski et al., 2015). Stemming from these results mitigation measures have been
55 applied to many roads to reduce wildlife-vehicle-collisions (Jackson & Griffin, 2000; Forman et
56 al., 2003; Grilo, Bissonette & Santos-Reis, 2009), allowing safe passage for animals, promoting
57 habitat connectivity, accessibility, and encouraging natural movements (Goosem, Izumi &
58 Turton, 2001; Bager & Rosa, 2010).

59

60 However, in developing countries, efforts to reduce wildlife mortality around main roads are
61 often hampered due to a lack of research, with other priorities usually dictated by the country's
62 socio-economic situation (Collinson et al., 2015a). For example, the collection of roadkill data
63 by dedicated research teams can be extremely costly due to the high sampling effort required
64 (Abra et al., 2018). This often limits the number of roadkill studies, and to date only a handful of
65 studies have focussed on roadkill in Africa (Collinson et al., 2015b), and yet these data are vital
66 in order to implement effective mitigation strategies (Gunson & Teixeira, 2015). This is
67 unfortunate because while Africa is incredibly rich in biodiversity (Mittermeier et al., 2011), it
68 also has the fastest growing human population (and associated infrastructure) in the world
69 (United Nations, 2015), which could have serious environmental impacts (Calvin et al., 2016).

70

71 One of the ways in which costly sampling strategies can be overcome is to adopt a citizen
72 science approach, whereby scientific data are collected by members of the public (Conrad &
73 Hilchey, 2011). For example, as a consequence of citizen science surveys used to monitor the
74 status of birds in the United Kingdom, the UK government introduced targets to reverse
75 population declines identified by the surveys (Gregory & van Strien, 2010; Greenwood, 2012).
76 While citizen science can be a powerful tool to collect scientific data, biases in survey effort in
77 such studies can hinder studies of variables such as roadkill rates. An alternative but

78 underutilised source of data on roadkill rates, that has the potential to offer more consistent and
79 measurable survey effort, is road patrols. Many highway agencies conduct regular patrols in
80 order to resolve any issues that could affect road users, so partnerships between researchers and
81 road patrol agencies offer significant potential for more effective data collection (Périquet et al.,
82 2018).

83
84 In this study, we used data collected by road patrol staff of the N3 Toll Concession (N3TC) to
85 explore the potential for partnering with road agencies to conduct roadkill studies. In 2011, a
86 partnership was established between the Endangered Wildlife Trust (EWT) and the N3 Toll
87 Concession (N3TC), an organisation that operates a 415 km of the N3 highway in South Africa,
88 with the aim of managing the impact of roads on wildlife. N3TC patrol staff began collecting
89 data on roadkill incidents, following training from the EWT. Exploratory analyses of the data
90 collected suggested that serval (*Leptailurus serval*) appear to be particularly susceptible to
91 roadkill on the N3 (Fig. 1).

92

93



94

95 Figure 1. Photograph of serval carcass taken by N3TC road patrol staff. Image credit: N3TC.

96

97 The serval is a medium-sized carnivore weighing 8-12 kg (Skinner & Chimimba, 2005). It
98 displays a preference for wetland habitats (Ramesh & Downs, 2015/2), the degradation and loss
99 of which constitute the principal threat to serval populations (Thiel, 2011). Rodents make up a
100 large proportion of serval diets, but they are also known to feed on other small prey such as
101 birds, reptiles, and insects (Ramesh & Downs, 2015). The species is listed as Least Concern on
102 the IUCN Red List of threatened species (Thiel, 2015), and Near Threatened on the South
103 African Red List (Ramesh et al., 2016), although serval numbers have declined throughout their
104 range (Ramesh & Downs, 2013). Home ranges can be between 8 and 38 km² (Geertsema, 1985;
105 Ramesh, Kalle & Downs, 2015), and population densities can vary from approximately 6-100
106 individuals per 100 km² (Ramesh & Downs, 2013; Loock et al., 2018). Servals display largely
107 crepuscular and nocturnal activity patterns (Thiel, 2011; Ramesh & Downs, 2013), but they can

108 also be active during the day (Geertsema, 1985). Although roadkill is thought to present a serious
109 threat to servals (Ramesh et al., 2016), there have been few attempts to quantify these threats or
110 identify associated risk factors.

111
112 We set out to use data collected by N3TC road patrol staff as part of routine road patrols to
113 identify which factors were related to serval roadkill counts on the N3. We modelled the
114 relationship between a set of environmental and anthropogenic variables in order to determine
115 which, if any, were related to serval roadkill counts, in order to inform the development of
116 potential strategies to mitigate serval roadkill.

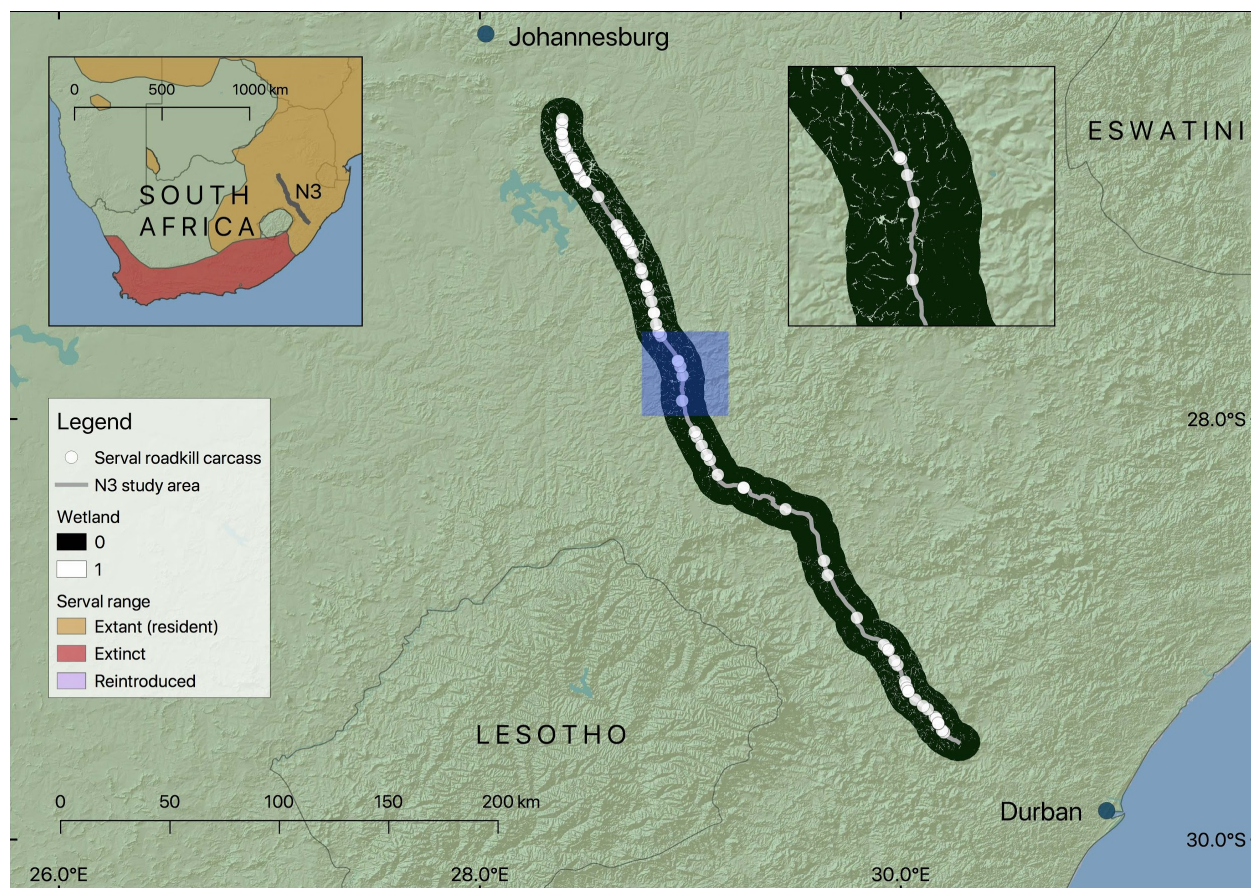
117

118 2. Materials & Methods.

119 2.1. Study site

120 The study focussed on a 410 km stretch of the N3 Toll Route (hereafter referred to as the N3), a
121 major road linking Johannesburg and Durban in South Africa (Fig. 2). It is of strategic
122 importance as it links the largest and third largest cities in the country, and it is a major route for
123 transporting goods from South Africa's largest port. The N3 is classified as a national route, the
124 highest category in the South African road network. Most segments of the national route network
125 are maintained by the South African National Roads Agency (SANRAL), but some are
126 maintained by provincial, local, or private road authorities. The N3 has been operated by N3TC
127 since 1999 (N3 Toll Concession, 2018). Annual traffic volume for the section of the road studied
128 was approximately 340,000 vehicles per km between 2014 and 2017, and the road is generally
129 two lanes wide in each direction, with a speed limits of either 100 or 120 km/h. Passing through
130 three of the eleven South African biomes (Savanna, Grassland and Indian Ocean Coastal Belt
131 (Mucina & Rutherford, 2006)), the N3's diverse range of habitats also includes urban landscapes,
132 communal land, and agricultural areas.

133



134

135 Figure 2. Map showing the location of the section of the N3 studied, serval roadkill carcass
136 locations recorded from 2014 to 2018, wetland within 10 km of the road, and serval range in
137 South Africa (adapted from Thiel (2015)). Inset map (location shown in blue) shows a closer
138 view of serval roadkill carcass locations in relation to wetlands.

139

140 2.2. Data collection

141 N3TC road patrol staff collected field data as part of their routine patrols of a 410 km stretch of
142 the N3 between Johannesburg and Durban (Fig. 2) every day between 01/01/2014 and
143 31/12/2017. The road patrol staff involved in the study undertook annual training with the
144 Endangered Wildlife Trust (EWT) in wildlife identification and the collection of roadkill data.
145 Patrols were driven four times a day (twice in each direction), with teams of two patrollers
146 allocated to six shorter sections to ensure the entire road was covered effectively. Patrols were
147 conducted at a speed of approximately 100 km/h.

148

149 Observers recorded the species of roadkill carcasses encountered, the date and time of the
150 observation, and the carcass location using the nearest route marker, which were located at 200
151 m intervals along the road. Carcasses were removed from the road to avoid recounts (Collinson
152 et al., 2014; Guinard, Prodon & Barbraud, 2015). For the current study we focussed on records
153 of serval, owls (order Strigiformes), and guineafowl (family Numididae) roadkill, as these were
154 among the most numerous. We classified roadkill observations collected between the 1st of
155 October to the 31st of March as occurring during the wet season, and roadkill collected from the
156 1st of April to the 31st of September as occurring during the dry season (Cook, Reason &
157 Hewitson, 2004).

158

159 In addition to roadkill data collected by the patrol teams, we also used existing data on habitat
160 type (Geoterraimage, 2015). We also obtained data from N3TC on traffic volumes and speed
161 limits, and the locations of infrastructure such as bridges and underpasses. We measured road
162 width at 410 randomly-generated locations using Copernicus Sentinel satellite imagery
163 (European Space Agency, 2017), measured in QGIS v3.0 (QGIS Development Team, 2018).
164 Road shapefiles were obtained from OpenStreetMap (OpenStreetMap contributors, 2017).

165

166 2.3. Data analysis

167 We conducted all analyses in R v3.5.0 (R Development Core Team, 2018), and all code and data
168 are publically available (Williams et al., 2018). We divided the N3 into 41 sampling units, each
169 with a length of 10 km, corresponding to the home range size of serval (Geertsema, 1985;
170 Bowland, 1990; Ramesh, Kalle & Downs, 2016). Within each sampling unit we summed the
171 total number of serval carcasses, which we used as the response variable in our models. All other
172 environmental and anthropogenic variables included in the models were used as potential
173 predictors (Table 1).

174

175 We developed several hypotheses that could explain the number of serval roadkills (Table 1).
176 First, we hypothesised that rodent abundance near the edges of roads could affect serval roadkill
177 (Boyle, Litzgus & Lesbarrères, 2017). We used the number of owl roadkill as a proxy for rodent
178 abundance since there is a well-established relationship between the abundance of owls and
179 rodents (Ansara, 2008; Bullock, Malan & Pretorius, 2011). Secondly, we hypothesised

180 guineafowl might be a food source for serval around road edges. We calculated the total number
181 of owl and guineafowl roadkill carcasses per sampling unit from the roadkill survey data. Since
182 serval prefer certain habitats, we also hypothesized that habitat type, in particular the amount of
183 available wetland (Ramesh & Downs, 2015/2), would affect serval roadkill. We estimated
184 habitat variables by first creating a buffer with a radius of 10 km around each sampling unit, and
185 used land cover data (Geoterraimage, 2015) to calculate the most common habitat type, and the
186 proportion of wetland within the buffer for each sampling unit. As carnivore home ranges are
187 typically larger in seasons with lower rainfall (Macdonald & Loveridge, 2010), we hypothesised
188 that serval roadkill counts would be higher during the dry season, as larger home ranges would
189 be associated with increased contact with roads.

190
191 There is evidence that anthropogenic variables such as traffic volume and speed can also
192 influence roadkill rates (Jaarsma, van Langevelde & Botma, 2006; Garriga et al., 2012). We thus
193 also included several anthropogenic predictor variables (Table 1). To generate anthropogenic
194 predictors we calculated the average speed limit, traffic volume, road width, and the total number
195 of road infrastructure points for each sampling unit. We used the variance inflation factor (VIF)
196 in package car (Fox & Weisberg, 2011) to check for colinearity among predictor variables, and
197 excluded variables with a VIF > 3 (Zuur et al., 2009).

198
199

200 Table 1. Hypotheses about the relationships between serval roadkill counts and predictor
 201 variables included in the models.

Type	Variable name	Variable description	Direction of association	Rationale
Environmental	Owls	Count of owl carcasses	+	Areas with a greater abundance of serval prey will support greater densities of serval and owls
	Guineafowl	Count of guineafowl carcasses	+	Guineafowl may be preyed upon by serval, increasing serval density
	Wetland	Proportion of 10 km buffer composed of wetland	+	Serval would be more abundant in areas rich in their preferred habitat
	Season	Wet or dry season	More in dry	Serval may range over larger distances when water is scarce, making them more vulnerable to WVCs
Anthropogenic	Traffic	Average number of vehicles per km	-	More vehicles present more opportunities WVCs
	Speed	Average speed limit (km/h)	-	Faster cars will be more difficult to avoid
	Road width	Average width of road (m)	-	Wider roads take longer to cross
	Infrastructure	Total number of infrastructure points such as bridges and underpasses	-	Bridges and underpasses may provide more opportunities for serval to cross roads safely

202

203

204 We modelled the relationship between serval roadkill counts and predictor variables by fitting a
 205 generalised linear mixed-effect model using the lme4 package (Bates et al., 2015). Exploratory
 206 analyses showed that negative binomial or Poisson distributions were the most plausible for our

207 dataset. The negative binomial distribution with log link outperformed other model structures
208 based on the Akaike Information Criterion (AIC) and visual inspection of a serval roadkill count
209 histogram comparing distributions (Figure S1), which was expected as the data were over-
210 dispersed with an excess of zeros. This model structure was therefore used for the remaining
211 analyses, following Valdivia et al. (2014).

212

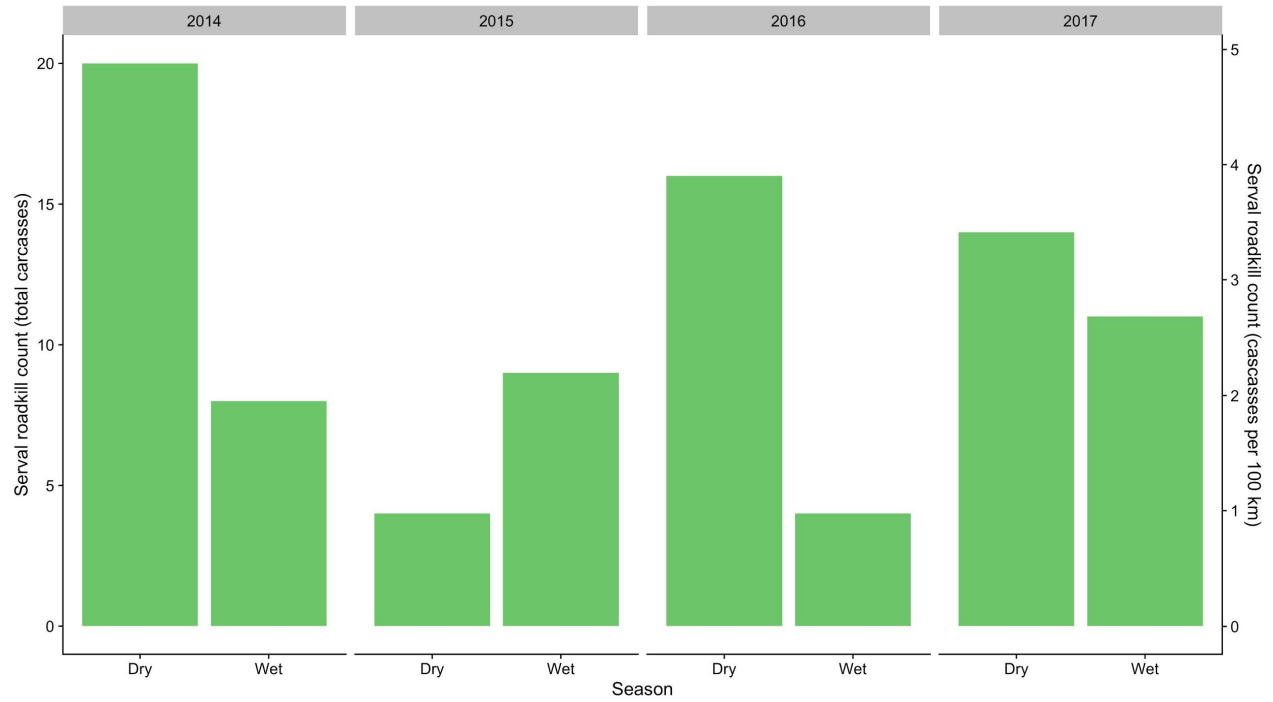
213 We included each of the predictor variables listed in Table 1 as fixed terms in the model. To
214 account for spatial autocorrelation we included sampling unit, nested within the most common
215 habitat type, as random terms in the model, including an offset to account for the length of the
216 sampling unit. We checked that there were no patterns in plots of fitted values against residuals
217 to validate the model (Figure S2), and confirmed that there were no patterns evident in plots of
218 predictor variables against residuals. We confirmed that spatial autocorrelation was not a
219 problem in our models by performing Mantel tests (Mantel, 1967) (Figure S3) and by inspecting
220 the sample variogram for residuals (Figure S4). Overdispersion did not appear to be a problem
221 for the final model ($\theta = 1.10$).

222

223 3. Results.

224 A total of 86 serval roadkill events were recorded at the study site between 01/01/2014 to
225 31/12/2017 (Fig. 3), which is equivalent to 5.24 carcasses/100 km/year. Our model showed
226 support for effects of season, wetland, and owls on serval roadkill counts (Fig. 4, Output S5).
227 The number of serval roadkill was greater in the dry season than the wet season, and in areas
228 with more wetland, and with higher owl roadkill counts (Fig. 5). There was no support for effects
229 of the number of guineafowl roadkill, vehicle speed limit, traffic volume, road width, or the
230 amount of road infrastructure on serval roadkill (Fig. 4, Output S5).

231

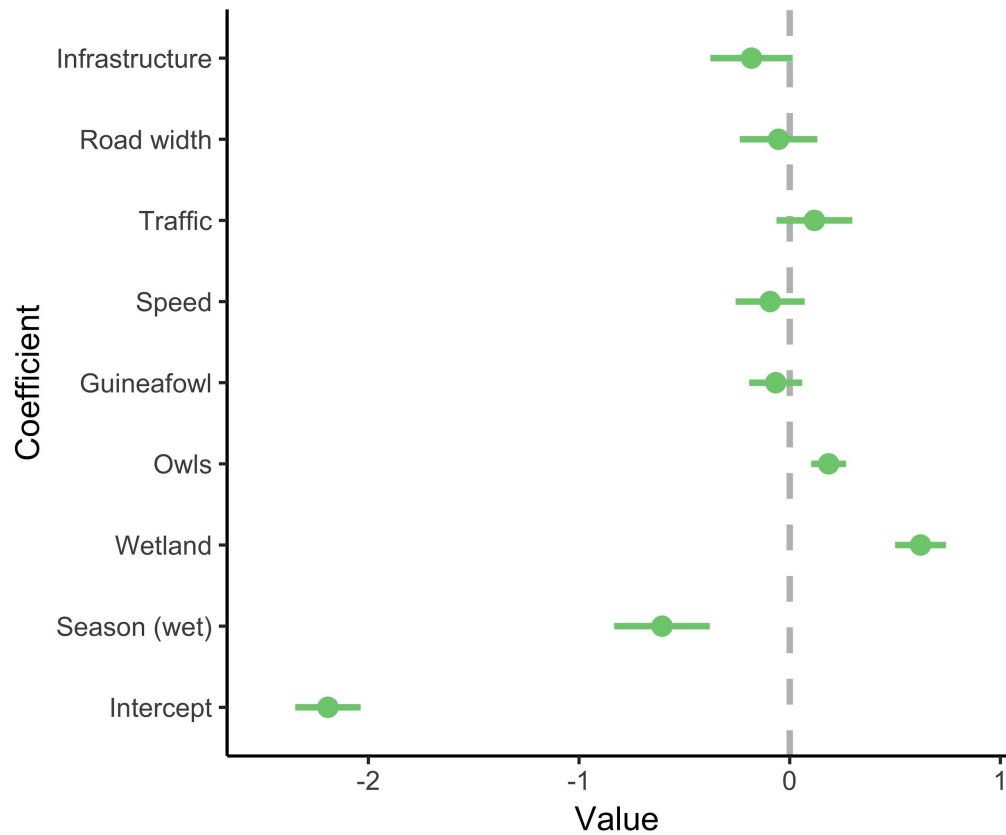


232

233 Figure 3. Total seasonal serval roadkill counts collected along a 410 km section of the N3 in

234 South Africa between 2014 and 2017.

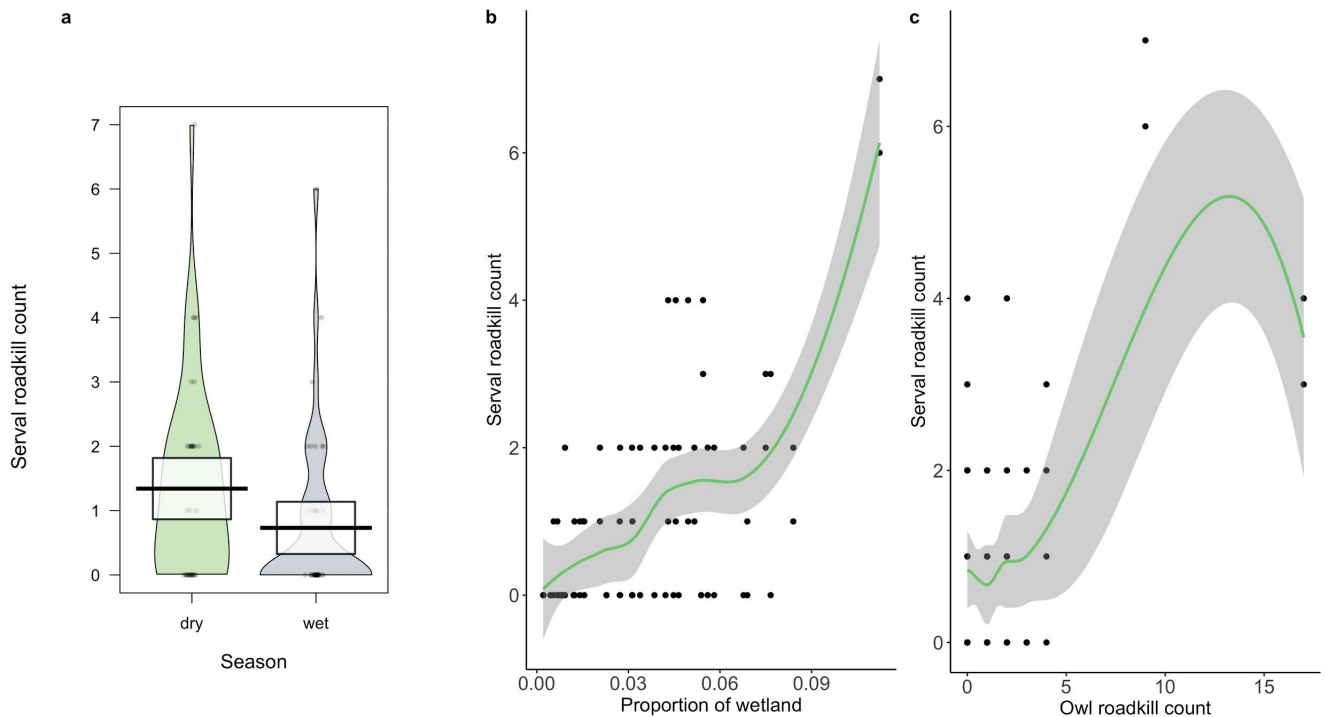
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236

237 Figure 4. Coefficient estimates showing the effect of predictor variables on serval roadkill
238 counts. Error bars represent 95% confidence intervals. We modelled roadkill counts using a
239 generalized linear mixed effect model with a negative binomial distribution and log link. The full
240 model summary is provided in Output S5.

241



242

243 Figure 5. The relationship between serval roadkill counts in each 10km sampling unit and a)
 244 season, b) proportion of wetland; and c) owl roadkill counts on the N3 between 2014 and 2017.

245 Boxes (Fig. 5a) and shaded areas (Fig. 5b and c) show 95% confidence intervals.

246

247 4. Discussion.

248 Our findings demonstrate that partnering with road patrol agencies can be a powerful strategy for
 249 collecting the necessary data to assess factors associated with roadkill rates, enabling targeting of
 250 mitigation strategies. Conducting twice-daily transects along both lanes of a 410 km stretch of a
 251 major highway solely to collect roadkill count data for research purposes would have been
 252 prohibitively costly, but by partnering with a road patrol agency we were able to gather the
 253 necessary data much more efficiently, and in a standardised manner (Collinson et al., 2014). We
 254 suggest that the development of partnerships between other road agencies and conservation or
 255 wildlife management organisations could also help to efficiently reduce roadkill rates, benefitting
 256 both drivers and biodiversity conservation and guide future mitigation measures (Abra et al.,
 257 2018).

258
259 We found support for the hypothesis that serval roadkill counts would be positively related to the
260 amount of wetland, which suggests that servals are being killed on roads located in their
261 preferred habitat (Ramesh & Downs, 2015/2), probably because they are more abundant in these
262 areas (D'Amico et al., 2015). Similar findings were found with the raccoon dog (*Nyctereutes*
263 *procyonoides viverrinus*) in Japan (Saeki & Macdonald, 2004), spotted turtles (*Clemmys guttata*)
264 and Blanding's turtles (*Emydoidea blandingii*) turtle in North America (Beaudry, deMaynadier &
265 Hunter, 2008), and owls in Portugal (Gomes et al., 2009). Our findings also support the
266 hypothesis that there would be a positive association between roadkill counts of serval and owls.
267 Areas with more owl roadkill may have a greater abundance of species that both owls and servals
268 prey upon (Vernon, 1972; Ramesh & Downs, 2015), which supports a greater abundance of both
269 owls and serval. Finally, we found support for the hypothesis that serval roadkill counts would
270 be higher in the dry season than the wet season, as carnivores tend to have larger home ranges
271 when water is scarce (Macdonald & Loveridge, 2010), which could put them at greater risk of
272 WVCs.

273
274 We found no support for any of our other hypotheses, including that serval roadkill counts would
275 be associated with guineafowl roadkill counts, or with anthropogenic variables such as traffic
276 volume, speed limit, road width, or the amount of road infrastructure. This suggests that serval
277 roadkill counts are influenced primarily by ecological drivers, and that this will vary between
278 species with different ecological requirements. As a result roadkill mitigation strategies may
279 need to be tailored to target species rather than relying on a more generalised approach.

280
281 We calculated an average of 5.24 serval carcasses/100 km/year (85 carcasses in total) on the N3,
282 but there are few studies with which this can be compared. Serval were among the most common
283 species recorded in a previous study on wildlife roadkill on the N3 (14.8% of 183 citizen
284 scientist records, and 1.5% of 209 road patrol records between 2011 and 2014) (Périquet et al.,
285 2018). Five of 17 collared serval were killed by snaring, hunting dogs, and WVCs in the
286 Drakensberg Midlands, South Africa (Ramesh, Kalle & Downs, 2016), but the breakdown of
287 these mortality sources is unclear. Serval roadkill has also been occasionally recorded elsewhere

288 in Africa, with two serval carcasses recorded in a study on roadkill in Uganda (Cibot et al.,
289 2015), but without data on survey effort this cannot be converted into a rate for comparison.

290
291 The rate of serval roadkill on the N3 is roughly in line with roadkill rates of other carnivores
292 such as spotted hyaena (*Crocuta crocuta*), common genet (*Genetta genetta*), and black-backed
293 jackal (*Canis mesomelas*) in the Tarangire–Manyara ecosystem, Tanzania (each at approximately
294 5 individuals/100km/year) (Kioko et al., 2015). Carnivore roadkill rates in Portugal were also
295 similar (5-6 individuals/100 km/year) for species including Egyptian mongoose (*Herpestes*
296 *ichneumon*), Eurasian badger (*Meles meles*), and common genet (Grilo, Bissonette & Santos-
297 Reis, 2009). Carnivore roadkill rates can, however, vary greatly, for example from 1 to 20
298 individuals/100 km/year for Western polecats (*Mustela putorius*) and red fox (*Vulpes vulpes*)
299 respectively (Grilo, Bissonette & Santos-Reis, 2009).

300
301 Without data on serval population sizes in the study area, or on other sources of mortality, it is
302 difficult to determine the impacts that WVCs may have on the population dynamics of the serval
303 population. Our findings nevertheless appear to support the suggestion by Ramesh et al. (2016)
304 that WVCs could be a major source of mortality for the species, which is also true for other
305 carnivores. Over 40% of Eurasian badgers in southwest England, for example, were thought to
306 be killed each year in WVCs (Clarke, White & Harris, 1998). WVCs accounted for 35% of
307 annual mortality of Florida panthers (*Puma concolor coryi*) in the USA (Taylor et al., 2002), 4-
308 33% of annual mortality of jaguars (*Panthera onca*) in the Atlantic Forest, Brazil (Cullen et al.,
309 2016), and 17% of annual mortality of Iberian lynx (*Felis pardina*) in southern Spain (Ferrerias et
310 al., 1992). Roadkill mitigation strategies could therefore contribute significantly conservation
311 efforts for some species, so identifying factors that are associated with high roadkill rates is a
312 useful first step. In this case study, one of the implications of our findings is that in order to
313 reduce serval roadkill, mitigation efforts should be focussed in areas rich in wetland.

314
315 One measure that could help to reduce serval roadkill near wetlands is installing a combination
316 of fencing and wildlife crossing structures (Clevenger, Chruszcz & Gunson, 2001). Wildlife
317 under- and overpasses are used extensively in Europe and America (Danielson & Hubbard, 1998;
318 Goosem et al., 2008; Bissonette & Rosa, 2012), and numerous studies have reported on the value

319 of fencing in reducing wildlife road mortalities (Clevenger, Chruszcz & Gunson, 2001; Jaeger &
320 Fahrig, 2004). This can be an effective solution (Seiler, 2005), but the tendency of serval to use
321 wildlife crossing structures has not yet been assessed. These structures are widely used by many
322 species (Clevenger, Chruszcz & Gunson, 2001), including carnivores (Grilo, Bissonette &
323 Santos-Reis, 2008; Andis, Huijser & Broberg, 2017), so it is likely that they would be used by
324 serval also, although we recommend testing this. We note, however, that care must be taken with
325 this approach, as fencing can also reduce gene flow and fragment habitats (Flesch et al., 2010).

326

327 Another important factor to consider when planning the location of mitigation measures such as
328 wildlife crossing structures and fencing is wildlife movement patterns. Movement patterns of
329 many species are often associated with drainage lines, topography, and habitat, and through
330 understanding the characteristics of target species, measures can be implemented at optimal sites
331 (Clevenger, Chruszcz & Gunson, 2001; Rytwinski et al., 2016). Correct tunnel design and
332 installation, as well as the presence of guiding fences are all factors to be considered when
333 designing wildlife passages (Danielson & Hubbard, 1998).

334

335 In addition to the above, research suggests that using these measures alongside other methods
336 such as implementing low speed zones, installing traffic-calming devices such as speed bumps
337 (Glista, DeVault & DeWoody, 2009), or using species-specific warning-signage placed in
338 strategic spots to alert road users to serval presence (Hardy, Lee & Al-Kaisy, 2006), may
339 increase the effectiveness of the mitigation measures. Our data, however, suggests that
340 anthropogenic variables such as traffic speed limits may not be an important predictor of serval
341 roadkill rates on the N3, so these measures may be less effective in this situation. Removing
342 roadside vegetation may also reduce the rate of serval and owl roadkill, as this could reduce the
343 abundance of rodent prey species (Ruiz-Capillas, Mata & Malo, 2015), although this could
344 impact the conservation of other species such as plants and small mammals. Keeping grass
345 verges trimmed, rather than removing vegetation entirely, may also be effective, by manipulating
346 the landscape of fear (Jacob & Brown, 2000) and causing wildlife to spend less time in verge
347 habitats without altering their abundance. A further potential issue that could impact roadkill
348 rates on the N3 is grain spilled on the road from agricultural vehicles (Ansara, 2008;
349 Gangadharan et al., 2017). If this is a common occurrence the increased food availability could

350 boost rodent abundance, drawing predators into the road and leading to increased WVCs
351 (Gangadharan et al., 2017). We suggest that this would be a worthwhile area of future study.

352
353 Although useful, there were some caveats to this approach. For example, road patrol teams did
354 not collect data that would have allowed us to model the number of missed detections (Shilling,
355 Perkins & Collinson, 2015). Furthermore, recording data on which patrol teams conducted which
356 patrols would have enabled us to control for inter-observer bias (Shilling, Perkins & Collinson,
357 2015). We were unable to record the age-sex class of carcasses, which would have been
358 interesting to include in the models (Pallares et al., 2015). Additional data such as the density of
359 servals and prey species at different sampling points along the length of the road would have
360 been incredibly useful in determining the impact on local populations. Finally, the speed at
361 which patrols were driven could bias roadkill detection towards larger species (Collinson et al.,
362 2014), so this method would not be suitable to collect roadkill counts for all species. Further
363 studies incorporating missed carcass detections, inter-observer bias, and population density of
364 target species would be worthwhile. Despite these limitations, this study demonstrated that even
365 the collection of very limited data such as the location, date, and species (or taxon) of roadkill
366 carcasses can help to inform wildlife management policy while minimising additional workload
367 burden on road patrol staff.

368

369 5. Conclusions.

370 Integrating roadkill data collection into existing road patrols can provide an efficient means of
371 collecting data to allow identification of factors associated with carnivore roadkill. Our findings
372 support the hypotheses that serval roadkill counts were higher in the wet season than the dry
373 season, and they were also higher in areas with more wetland and in areas with higher owl
374 roadkill counts. Anthropogenic factors such as traffic volume, speed limit, and the amount of
375 road infrastructure did not influence serval roadkill counts. We suggest that efforts to mitigate
376 serval roadkill, such as installing wildlife crossing structures in combination with fencing, should
377 be targeted at wetlands, but this must be tailored to the ecological requirements of target species.

378

379 6. Acknowledgements.

380 We are extremely grateful to N3TC staff for collecting the field data and providing additional
381 data on traffic volumes and speed limits.

382

383 7. References.

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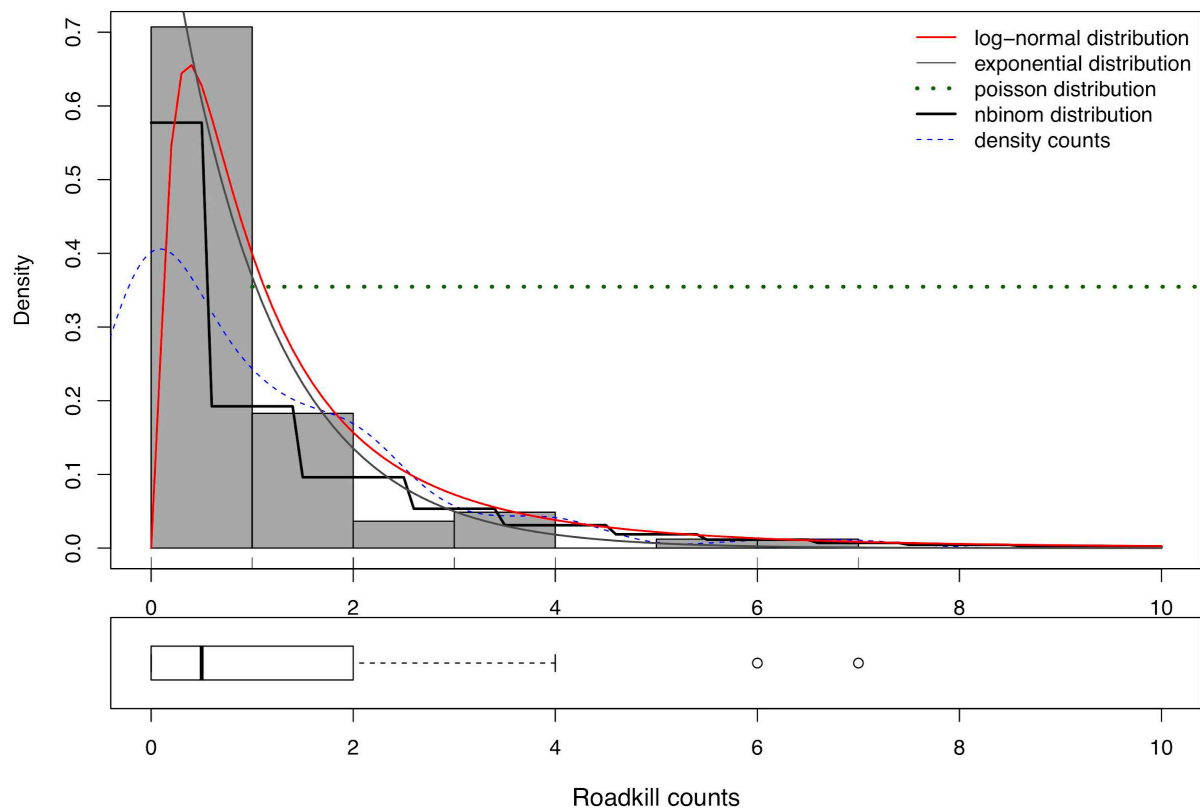
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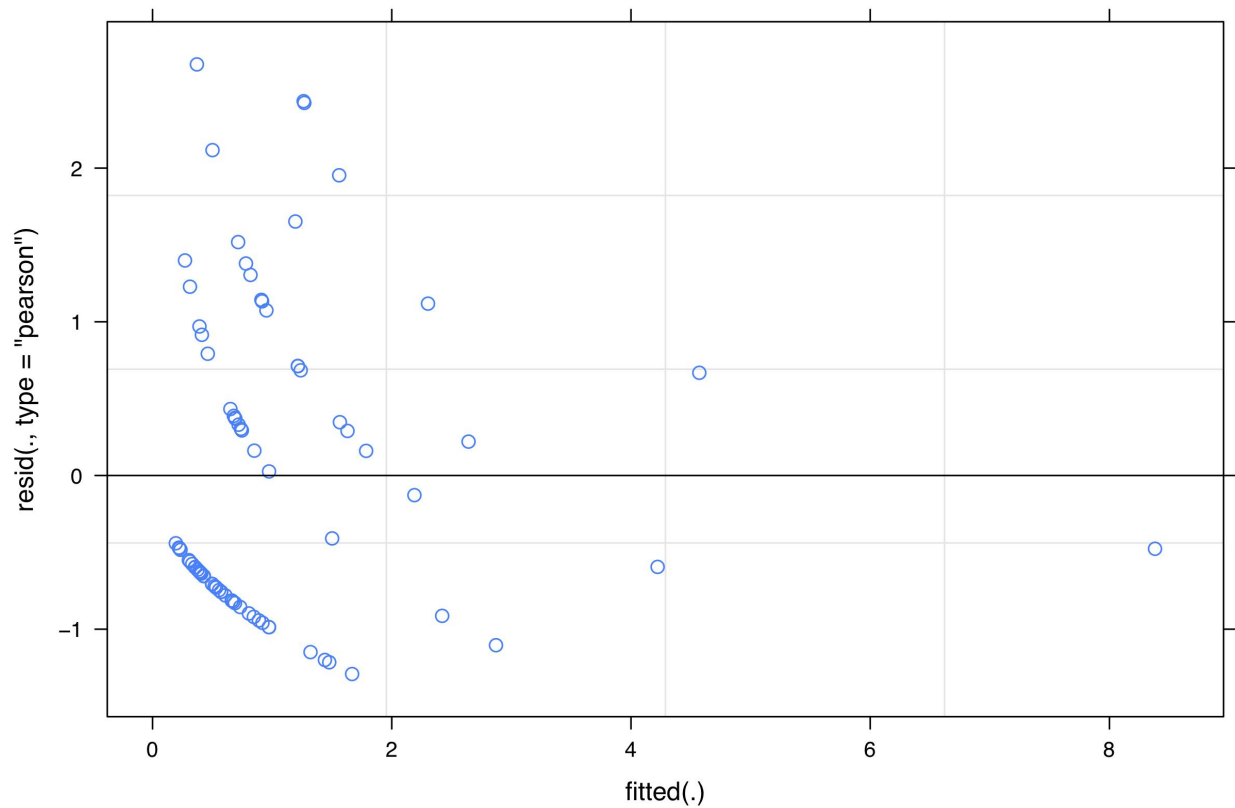
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620 8. Supplementary Information.

621 All data and code used to perform the analyses underpinning this paper are publically available
622 in Williams S., Collinson W., Patterson-Abrolat C., Marneweck DG., H. Swanepoel L. 2018.
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626
627 Figure S1. Histogram showing distribution of serval roadkill counts in relation to a range of
628 distributions.
629



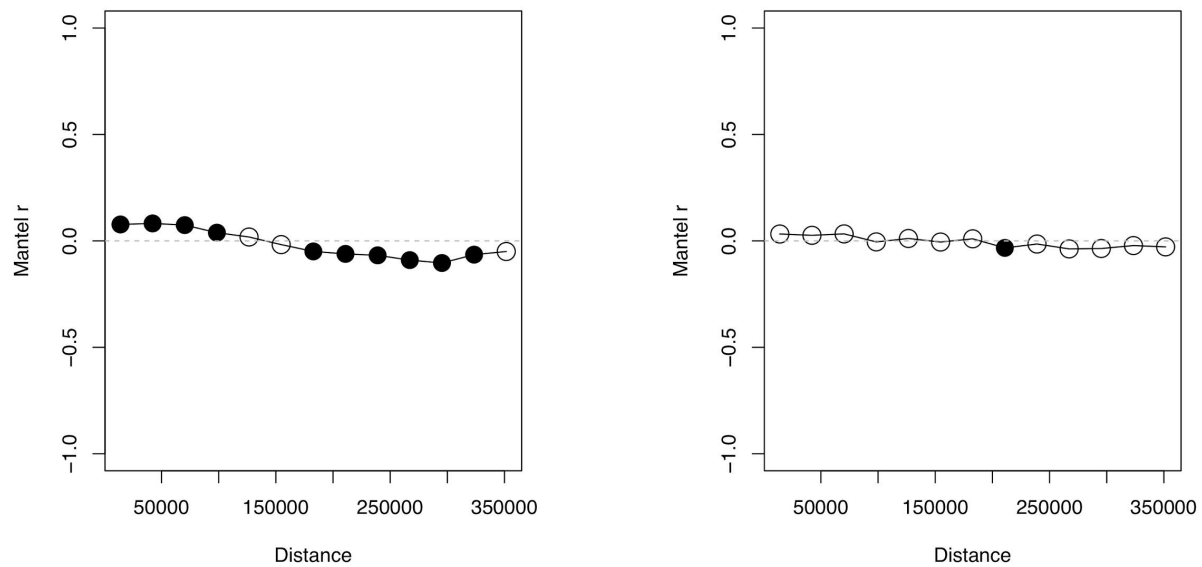
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631 Figure S2. Scatter plot of fitted vs residual values for the full generalized linear mixed model

632 with negative binomial distribution.

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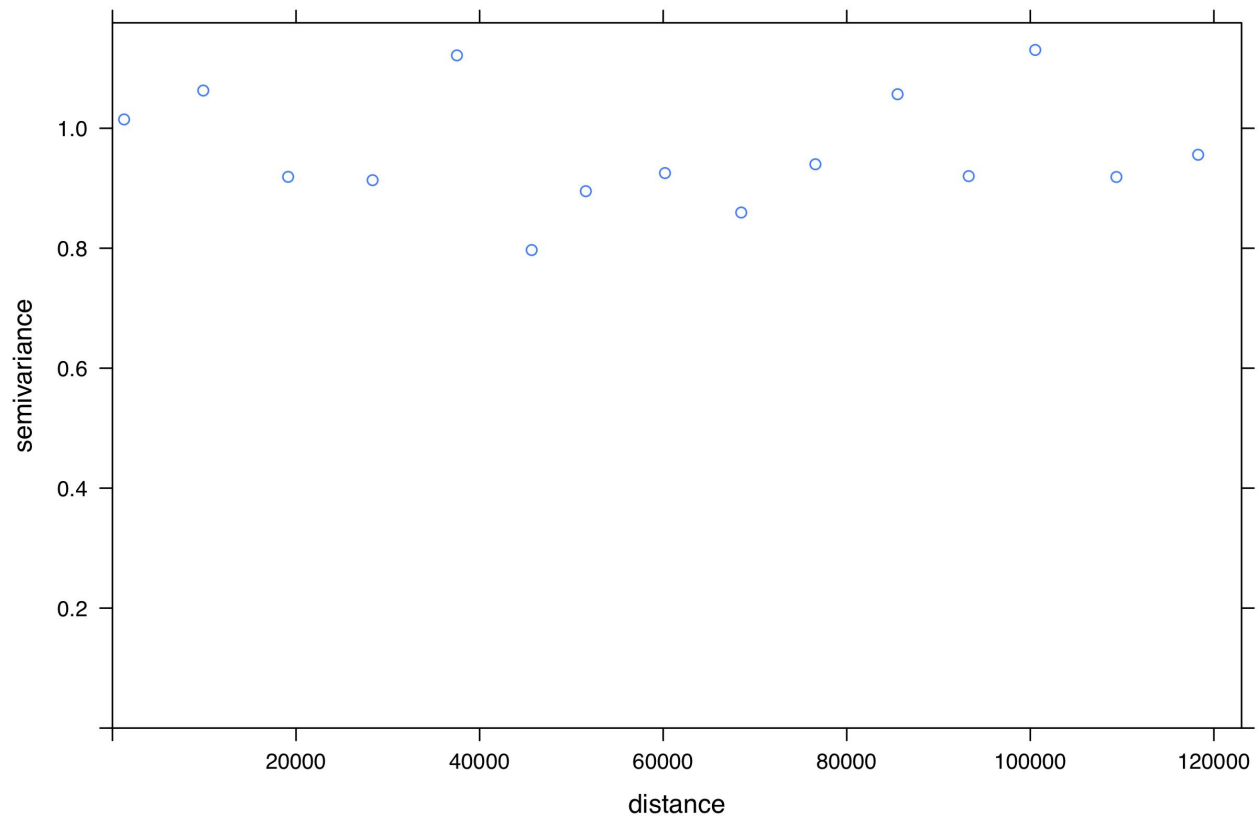
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636 Figure S3. Mantel tests for the full generalized linear mixed model with negative binomial
637 distribution.

638

639

640



641

642 Figure S4. Variogram from for residuals for the full generalized linear mixed model with
643 negative binomial distribution.

644

645

646 Output S5. Summary of full generalized linear mixed model with negative binomial distribution.

647

648 Generalized linear mixed model fit by maximum likelihood

649 (Laplace

650 Approximation) [glmerMod]

651 Family: Negative Binomial(25923.52) (log)

652 Formula:

653 roadkill ~ season + scale(wetland) + scale(owls) +

654 scale(guineafowl) +

655 scale(speed) + scale(traffic) + scale(roadWidth) +

656 scale(infrastructure) +

657 offset(log(LogArea)) + (1 | habitat/section)

658 Data: RKdata2

659

660 AIC BIC logLik deviance df.resid

661 213.6 242.4 -94.8 189.6 70

662

663 Scaled residuals:

664 Min 1Q Median 3Q Max

665 -1.2916 -0.7416 -0.4805 0.6802 2.6747

666

667 Random effects:

668 Groups Name Variance Std.Dev.

669 section:habitat (Intercept) 2.435e-11 4.934e-06

670 habitat (Intercept) 4.208e-12 2.051e-06

671 Number of obs: 82, groups: section:habitat, 41; habitat, 3

672

673 Fixed effects:

674 Estimate Std. Error z value Pr(>|z|)

675 (Intercept) -2.19221 0.15516 -14.129 < 2e-16 ***

676 seasonwet -0.60616 0.22678 -2.673 0.00752 **

677 scale(wetland) 0.62091 0.12052 5.152 2.58e-07 ***

```

678 scale(owls)          0.18479    0.08259    2.237    0.02527 *
679 scale(guineafowl)   -0.06648    0.12584   -0.528    0.59732
680 scale(speed)        -0.09286    0.16380   -0.567    0.57080
681 scale(traffic)       0.11736    0.17973    0.653    0.51378
682 scale(roadWidth)    -0.05302    0.18414   -0.288    0.77341
683 scale(infrastructure) -0.18124    0.19522   -0.928    0.35322
684 ---
685 Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
686
687 Correlation of Fixed Effects:
688      (Intr) sesnwt scl(wt) scl(wl) scl(g) scl(s) scl(t)
689 scl(W)
690 seasonwet  -0.516
691 scal(wtlnd) -0.352  0.000
692 scale(owls) -0.120  0.000 -0.358
693 scal(gnfwl)  0.040  0.000 -0.162  0.146
694 scale(sped)  0.008 -0.001 -0.373 -0.073 -0.298
695 scal(trffc) -0.136  0.000  0.241  0.251  0.042 -0.079
696 scl(rdWdth)  0.110  0.001  0.011 -0.218  0.429 -0.306 -0.216
697 scl(nfrstr)  0.112  0.000 -0.384  0.060 -0.008  0.140 -0.645 -
698 0.202
699

```