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Additional Information

1 **Movement patterns of forest elephants (*Loxodonta cyclotis* Matschie,**
2 **1900) in the Odzala-Kokoua National Park, Republic of Congo.**

3

4 **Movement patterns of forest elephants in the Odzala-Kokoua National Park.**

5

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14 **Abstract**

15 African forest elephants (*Loxodonta cyclotis* Matschie, 1900) are ecological engineers that
16 play a fundamental role in vegetation dynamics. The species is of immediate conservation concern,
17 yet it is relatively under-studied. To narrow this knowledge gap, we studied the drivers of daily
18 movement patterns (linear displacements) of forest elephants —characterised by a set of
19 geographical, meteorological and anthropogenic variables— in the Odzala-Kokoua National Park,
20 Republic of Congo. Explicitly, we used conditional random forest to model and disentangle the main
21 environmental factors governing the displacements of six forest elephants, fitted with GPS-collars
22 and tracked over 16 months. Results indicated that females moved further distances than males,
23 while the presence of roads or human settlements disrupted elephant behaviour resulting in faster
24 displacements. Forest elephants moved faster along water courses and through forest with understory
25 dominated by Marantaceae forests and bais, but moved slower in savannahs. Finally, flood-prone
26 areas —described by elevation and accumulated precipitation— and higher temperatures prevented
27 longer displacements. We expect these results to improve the knowledge on the species movements
28 through different habitats, which would benefit its conservation management.

29

30 **Key words:** Central Africa; distance to road; elephant gender; linear displacements; GPS telemetry;
31 seasonal behaviour.

32

33 1 INTRODUCTION

34 Forest elephant (*Loxodonta cyclotis* Matschie, 1900) movement and feeding behaviour
35 (pollarding, peeling and trampling) modify forest structure, controlling the availability of resources
36 for other organisms (Poulsen et al., 2018), and maintaining habitats with open canopy such as bais —
37 forest clearings— (Fishlock, 2010; Metsio Sienne, Buchwald, & Wittemyer, 2014; Turkalo, 2013). A
38 reduction in the abundance of forest elephants can alter plant communities and ecosystem function
39 (Poulsen et al., 2018), which may have implications for the entire ecosystem due to cascading
40 effects. Central Africa holds the main part of African forest elephant populations, although they are
41 currently enduring a pronounced increase in poaching activities, particularly in Gabon and Congo,
42 where the bulk of the entire population exist (Breuer, Maisels, & Fishlock, 2016; Gobush, Mutayoba,
43 & Wasser, 2008; Maisels et al., 2013; Poulsen et al., 2017), but also in eastern Democratic Republic
44 of Congo, south-eastern Cameroon and south-western Central African Republic (Wasser et al.,
45 2015). From 2006 until the present, a severe decline in elephant numbers has been detected due to a
46 surge in ivory poaching to the extent that the increasing human-elephant conflict and habitat loss
47 have been overshadowed in importance (Convention on International Trade in Endangered Species
48 of Wild Fauna and Flora - CITES, 2016). The Republic of Congo still hosts large populations of
49 forest elephants, although there is concern there regarding the illegal ivory trade (Milliken, Burn,
50 Underwood, & Sangalakula, 2013; Wittemyer et al., 2014). Nonetheless, the estimated forest
51 elephant population in Odzala-Kokoua National Park (hereafter OKNP) was recorded to be at about
52 14,000 individuals (1.0 elephant per km²) (Blake et al., 2007) and it is still one of the highest
53 densities for the species.

54 Mammal species tend to suffer from a higher risk of extinction when they have higher body
55 mass and larger ranges, as well as when their habitat is more fragmented (Crooks et al., 2017;
56 Gonzalez-Voyer, González-Suárez, Vilà, & Revilla, 2016; Ripple et al., 2016). Thus, aside from the

57 commercial interest of their ivory and other anthropogenic side-effects, elephants are intrinsically at
58 greater risk of extinction, and the study of their range, movement patterns and linear displacement
59 throughout different landscapes are of special concern for their effective conservation. Information
60 about mammal movement and migrations is crucial for conservation plans, but is often scarce
61 (Bohrer, Beck, Ngene, Skidmore, & Douglas-Hamilton, 2014; Harris, Thirgood, Hopcraft, Croomsigt,
62 & Berger, 2009). Apart from the distribution of food and other key resources (Johnson, Kays,
63 Blackwell, & Macdonald, 2002), mammal movement patterns are mainly determined by interactions
64 between individual characteristics (e.g., maturity) and weather (Elliot, Cushman, Loveridge, Mtare,
65 & Macdonald, 2014). In order to implement effective conservation and management plans, a greater
66 understanding is required about elephant movements across large areas and over long periods
67 (Breuer et al., 2016; Loarie, Aarde, & Pimm, 2009). Both the distribution and movement patterns of
68 African Savannah elephants (*Loxodonta africana* Blumenbach, 1797) have been related to land cover
69 type (Loarie et al., 2009; Young, Ferreira, & van Aarde, 2009), weather (i.e., temperature and
70 rainfall) (Birkett, Vanak, Muggeo, Ferreira, & Slotow, 2012; Bohrer et al., 2014), as well as
71 elevation, slope, and the presence of water bodies (natural or artificial)(Graham, Douglas-Hamilton,
72 Adams, & Lee, 2009). These natural features interact with human infrastructures such as fences or
73 human settlements (Buij et al., 2007; Loarie et al., 2009), which shape elephant movement patterns.
74 Furthermore, spatial movements were shown to influence survival and reproduction rates in
75 savannah elephants (Goldenberg, Douglas-Hamilton, & Wittemyer, 2018). But, while determinants
76 of elephant distribution and movement patterns have been extensively studied for Savannah
77 elephants (Bohrer et al., 2014; Buij et al., 2007; de Knecht et al., 2011; S. Z. Goldenberg, Douglas-
78 Hamilton, Daballen, & Wittemyer, 2016; Shifra Z. Goldenberg et al., 2018; Young et al., 2009),
79 relatively little is known about the ecology of the forest elephant largely due to the difficulties of
80 undertaking direct observations of them in rainforests.

81 Previous studies suggest that forest elephants prefer to move at night (Kuwong, 2014; Turkalo,
82 Wrege, & Wittemyer, 2013), probably conditioned by the schedules of human activities, and that
83 seasonal displacements of forest elephants are a result of the distribution and fruiting patterns of diet
84 tree species (Blake, 2002; Mills et al., 2018; Short, 1983; White, 1992). In addition, elephants are
85 known to use baits for access to minerals (Metsio Sienne et al., 2014), while these sites also function
86 as social arenas where elephants exhibit aggregations (Fishlock, 2010; Turkalo et al., 2013). It is also
87 known that elephants benefit from herbaceous vegetation in recently logged forests (Clark, Poulsen,
88 Malonga, & Elkan, 2009; Stokes et al., 2010) and positively select for this vegetation type during
89 specific seasons (Mills et al., 2018). By contrast, non-surveyed roads outside protected areas —
90 which are not patrolled to control hunting or poaching— are formidable barriers to elephant
91 displacements (Blake et al., 2008; Laurance et al., 2006). Finally, it has been reported that elephant
92 abundance and distribution varies seasonally, depending on the water-forest matrix, the forest-
93 grassland matrix, and the distribution and phenology of fruit trees (Blake, 2002; Mills et al., 2018;
94 Schuttler, Blake, & Eggert, 2012).

95 The goal of this study is to disentangle the main environmental drivers governing the daily linear
96 displacements of forest elephants in the OKNP. Specifically, we hypothesise that forest elephants
97 stay near areas of high resource abundance (mainly food and shelter), away from human
98 disturbances, and perform long-distance seasonal displacements to satisfy their ecological and
99 behavioural requirements; thus a suite of interactions between natural and anthropogenic variables
100 drive their displacements and whether they stay put or migrate. To test this hypothesis we: i)
101 employed GPS collars to collect information on OKNP elephant spatial movements, ii) gathered data
102 on geographic (e.g., elevation or slope), meteorological (e.g., rainfall and temperature) and
103 anthropogenic (e.g., distance to villages and roads) variables for the study period, iii) and used
104 Conditional Random Forest (CRFs, Carolin Strobl, Boulesteix, Zeileis, & Hothorn, 2007) to
105 determine the most relevant variables influencing elephant daily displacements.

106

107 2 MATERIALS AND METHODS

108 2.1 Study area

109 The OKNP (Figure 1; 00°23'-01°10'N, 014°39'-015°14'E) is located 850 km north of
110 Brazzaville, Republic of Congo, and is geographically placed between the Batéké Plateau and the
111 northern Congo forests (Mbete, Ngokaka, Ntounta, & Vouidibio, 2010). The highest elevations are
112 found in the W and NE of the park, while the main basin that drains the park flows towards the SE.
113 Of its total area of 13,600 km², six percent of the park surface is covered by savannah (Aveling &
114 Froment, 2001). Nevertheless, this area is described as a forest-savannah mosaic, with dense gallery
115 forest along the deeper river valleys, and savannah areas dominated by grass species (family
116 Graminae), with scattered fire-resistant shrubs (Bermejo, 1999). Today, it is the oldest and largest
117 protected area in the country and is a Biosphere Reserve. The remainder of the park is composed of
118 Congolian evergreen forests and semi-deciduous transitional forest, Marantaceae forest, swamp
119 forests and secondary forests found along roads and abandoned settlements and clearings. At OKNP,
120 agriculture is forbidden, although some subsistence farming (manioc, corn, peanut) takes place on its
121 outskirts, triggering some human-elephant conflicts in the Park and its surroundings (Boukoulou et
122 al., 2012). Some tourist activity exists in the central part of the park, mainly between the savannah of
123 Mboko and the Lango bai and, as such, there are some buildings in this area: hangars, a small dock, a
124 small airstrip and two groups of accommodation buildings where most of the tourist activities occur.

125

126 [Figure 1 here]

127

128 2.2 Elephant movements data collection

129 We immobilized eight adult elephants and fitted them with GPS collars (African Wildlife
130 Tracking) following methods in Blake, Douglas-Hamilton, & Karesh (2001), between November and
131 December 2014. All International and Congolese guidelines for the care and use of animals were
132 followed. Collar #9 stopped working before being fitted and collar #2 (Female 2) apparently stopped
133 working a few days after the capture; consequently, this elephant was not considered in the analyses.
134 The remaining six collars worked properly, taking up to one fix per hour. From 22 November 2014
135 until 11 April 2016, a database of 14,780 fixes was obtained for the six elephants, which roamed
136 across the southern part of the park (Figure 1). On average, collars failed to transmit 47.6% of fixes,
137 resulting in a mean of 6 (3–13) fixes per day (see Figure B1 in supplementary material for a
138 graphical depiction about the distribution of fixes per month). The failure of data transmission was
139 exacerbated under dense forest canopy; thus, the longer an individual spent under the forest canopy
140 or other types of dense land cover, the fewer the fixes collected (D'Eonet al., 2002). This situation
141 partly determined the selection of the modelling technique used in this study (see section 2.4). In
142 general, the temporal distribution of fixes and intervals between them show marked differences
143 between different tracking devices and consequently render an uneven temporal distribution (Loarie
144 et al., 2009). This deficiency is usually addressed by removing the most inaccurate fixes and by
145 averaging the positions to one fix at a standard time step (e.g., every 8 or 24h de Knegt et al., 2011;
146 Loarie et al., 2009). However because we used only one type of tracking device, we did not remove
147 any locations to avoid losing data for forest and dense land cover types, which were of particular
148 interest. Nonetheless, we standardised the location of each elephant in 24h time steps to calculate the
149 daily linear displacement (km/24h). We averaged the values of the coordinates (X, Y) of the fixes
150 collected every 24h. Environmental variables were assigned at the site of mean 24h fixes, whereas
151 the daily linear displacement was calculated by summing half the distance from this site to both the
152 previous and subsequent locations of mean 24h fixes respectively.

153

154 2.3 Environmental covariates

155 The OKNP is characterised by an annual rainfall of 1,200 mm, spread over two rainy and two
156 dry seasons (Mbeté et al., 2010). Therefore, we assigned elephant locations to: Short Dry (Dec–Feb),
157 Short Rainy (Mar–May), Long Dry (Jun–Aug), and Long Rainy (Sep–Nov) seasons (see Figure B1
158 in supplementary material for a graphical depiction about the distribution of fixes per season). We
159 obtained daily rainfall and temperature data from the GSOD (Global Surface Summary of the Day)
160 website maintained by the NOAA (National Oceanic and Atmospheric Administration) within the
161 United States Department of Commerce. Four meteorological stations surrounding the OKNP were
162 selected (i.e., the meteorological stations located at Kellé, Makoua, Ouésso and Souanké). The
163 meteorological variables used were the maximum, minimum and mean temperature (Max., Min. and
164 Mean, respectively), precipitation, as well as the moving average (MA) for the 30 preceding days for
165 these four variables, in order to better describe the two rainy and dry seasons (Table 1). These
166 variables were calculated for each meteorological station and were interpolated at each mean 24h fix.
167 However, the reliability of the values collected each station for the mean 24h fix is inversely
168 proportional to the distance between them and the mean 24h fix. Therefore, the interpolation was
169 inversely weighted by the distance of the mean 24h fix to each meteorological station.

170

171 [Table 1 here]

172

173 We retrieved the Digital Elevation Model from the web site of the Consortium for Spatial
174 Information (CGIAR-CSI) and used it to infer the slope with ESRI® ArcMap™ 10.3.0.4322. The
175 land cover type and complementary geospatial data (roads, rivers, villages and territorial boundaries)
176 were obtained from the Ministry of Forestry and Environment (2011). Specifically, data related to

177 vegetation classes were derived from a satellite image (Landsat) and consisted of 15 different cover
178 types including forested areas of different canopy cover (e.g., savannah, dense closed forest, etc.),
179 water and flooded areas (e.g., swamp forest) and areas covered by clouds, which were eventually
180 removed from final analyses. The land cover type assigned to each elephant location was then found
181 at the coordinates of the mean 24 h fix. In the end, forest elephants travelled across ten out of the
182 thirteen different land cover types (Figure 1). The distance of each daily observation to the nearest
183 river was used as a proxy for water resource availability, and was also calculated with ESRI®
184 ArcMap™ 10.3.0.4322. Human footprint was characterised by means of distance to villages,
185 distance to the Lango lodge, and distance to main roads, which were calculated in the same way as
186 the distance to rivers. Finally, we also considered the sex of each individual in the analysis. A
187 graphical depiction of the correlations among variables can be found in the Appendix A of
188 supplementary material.

189

190 2.4 Data analysis with conditional random forests

191 Gathering information on either natural or artificial features typically leads to datasets
192 encompassing a mixture of continuous and categorical variables sampled at different scales (i.e.,
193 varying accuracies and different categorical scales; Table 1). Therefore, the relationship between the
194 daily linear displacement by the forest elephants and the environmental predictors (natural or
195 anthropogenic) were investigated with Conditional Random Forests (CRFs, Strobl et al., 2007).
196 CRFs is an ensemble machine learning technique based on the aggregation of a large set of
197 conditional inference trees (Hothorn, Hornik, & Zeileis, 2006), which has already been used to
198 determine the most relevant drivers of extinction risk in African mammals (Di Marco et al., 2014).
199 CRFs resolve the bias in the implementation of the original Random Forests (RFs) (Breiman, 2001)
200 towards variables with higher accuracy (e.g., distance to rivers) over those with a low number of

201 categories (e.g., elephant sex), balancing the role of each variable, regardless of the number of
202 unique values or categories (Strobl et al., 2007). Consequently, CRFs allow a reliable inference of
203 linear and non-linear effects and the relative importance of the environmental variables (Strobl,
204 Hothorn, & Zeileis, 2009).

205 We developed the CRFs in *R* (R Core Team, 2015) with the *party* package (Strobl et al., 2009).
206 Following previous studies, we were more interested in general patterns than in the behaviours of
207 any single elephant (Loarie et al., 2009). Therefore, data were pooled and, prior to model training,
208 the linear displacement was log transformed to reduce the influence of any particular displacement
209 (i.e., principally to reduce the potential influence of female 4 – F4). Nevertheless, to balance the
210 influence of each individual in the model predictions, the input data were inversely weighted by their
211 proportion of fixes over the entire dataset. Unlike previous studies comparing decision-tree induction
212 methodologies (Muñoz-Mas, Fukuda, Vezza, & Martínez-Capel, 2016), we did not consider
213 beforehand the correlation between variables because CRFs correct the bias towards groups of
214 correlated variables present in standard RFs (Strobl et al., 2007). As recommended in a number of
215 publications (Arlot & Celisse, 2010), we performed a stratified fivefold cross-validation (i.e., each
216 fold presented a similar proportion of cases per individual as the complete dataset) to search for the
217 most general model using the backward variable elimination approach (May, Dandy, & Maier,
218 2011). The fivefold cross-validation was repeated five times (i.e., 25 models in total) (Muñoz-Mas et
219 al., 2016) and the loss function corresponded to the weighted Mean Squared Error (MSE) to balance
220 the influence of the different number of fixes per elephant. As a common practice (Muñoz-Mas et al.,
221 2016), once the optimal parameters and variables were obtained, a single case-weighted CRFs was
222 calculated without cross-validation and employing the optimal variable set. Using the final model,
223 we evaluated the coefficient of determination (r^2), and the modelled effects and importance of the
224 variables.

225 The relationship between the most relevant input variables and the daily linear displacement by
226 forest elephants was graphically characterised with partial dependence plots (Friedman, 2001)
227 adapting the code implemented in the package *randomForest* (Liaw & Wiener, 2002). Partial
228 dependence plots are based on the predictions rendered by the CRFs obtained after substituting, one
229 at a time and sequentially, the inspected variable by the different values of the variable (Muñoz-Mas,
230 Fukuda, Pórtoles, & Martínez-Capel, 2018). Then, the resulting predictions are used to depict the
231 effect of the inspected variable over the response variable (i.e., mean effect and/or other statistics)
232 accounting for the effects of the other variables within the model by averaging their effects.
233 Consequently, partial dependence plots are a useful way to visualise the marginal effect of the target
234 variable on the response variable (Muñoz-Mas et al., 2016). Finally, the importance of the variables
235 was examined by employing the function implemented by the *party* package (Strobl et al., 2009).

236

237 3 RESULTS

238 The average linear displacement per elephant (i.e., distance between standardised/mean 24h
239 fixes) was 1.87 km per day (1.65 ± 1.25 for males and 2.05 ± 1.82 for females). Female 4 showed the
240 longest displacement per day (15.4 km, during the short dry season of 2016), while female 3 showed
241 the shortest (~ 0.0 km, during the short dry season of 2016). Per season, the smallest displacements
242 (i.e., distance between standardised/mean 24h fixes) were close to zero both for males and females.
243 Conversely, the largest distances travelled varied. The longest displacements occurred during the
244 short dry season (i.e., 11.3 and 15.4 for males and females, respectively) whereas these values were
245 smaller during the other seasons and did not show a clear pattern between seasons and sexes (See
246 Appendix B for a graphical depiction of the linear displacements per month performed by each
247 elephant).

248 The backward variable elimination stopped after removing six of the seventeen variables. The
249 final variable set included nine continuous variables: the moving average of temperature and its
250 maximum and minimum and those of precipitation (i.e., Temp. MA, Max. Temp. MA, Min. Temp.
251 MA and Precipitation MA), distance to rivers and elevation and distance to villages, lodge and roads.
252 It also included two categorical variables: land cover type and sex. The final model explained 61
253 percent of the variance (Figure 2).

254

255 [Figure 2 here]

256

257 Although interactions between variables generated a range of predictions for each permuted
258 value, in general elephants moved furthest in the vicinity of main roads (Figure 3). Distance to roads
259 generated the largest differences in daily displacement (i.e., from approximately 0.7 to 1.1), which is
260 indicative of the importance of this variable (see Figure 5). Daily linear displacements were higher
261 for fixes located near to the Lango lodge (Figure 3), which were also collected at intermediate
262 distance to villages (i.e., between ca. 25 and 30 km); the abrupt increase in the distances moved at
263 fixes located more than 36 km from villages correspond to a very large displacement undertaken by
264 individual Female 4. Elevation indicated that the daily linear displacement by the elephants
265 decreased in flood-prone areas. Daily distance increased markedly from the lowest elevation (i.e.,
266 zero level) up to 425 m a.s.l., while above this elevation, there was no effect on linear displacement.
267 Forest elephants travelled almost irrespectively of the value of the moving average of the maximum
268 temperature (Max. Temp. MA) up to 31.3 °C. Above this temperature their movement decreased
269 abruptly. The moving average of the temperature (i.e., Temp. MA) had similar effect (elephants
270 displaced longer distances at lower temperatures), although no abrupt changes were present in this
271 case. On the contrary, the moving average of the minimum temperature (i.e., Min. Temp. MA)

272 presented an unclear pattern, although, overall, it can be considered to influence negatively elephant
273 movements. Finally, the moving average of precipitation (Precipitation MA) had almost a negative
274 linear effect so elephants moved shorter distances during periods of heavy rainfall.

275

276 [Figure 3 here]

277

278 The categorical variables selected indicated that forest elephants moved faster when moving
279 through Marantaceae forest, bays and land cover patches classified as water, mainly flooded river
280 flooding areas (Figure 4; left panel). Conversely, they moved more slowly in the remaining habitat
281 types, particularly in savannahs, which is interpreted as a preference for these habitats: elephants that
282 use savannahs may spend much time moving through slowly and grazing at a single location.
283 Finally, with regard to elephant gender, female elephants travelled longer daily distances than males
284 do (Figure 4; right panel).

285

286 [Figure 4 here]

287

288 The variable importance function indicated that sex and distance to roads and to lodge were
289 the most important variables determining the daily linear displacement by forest elephants, whereas
290 some meteorological variables (i.e., precipitation MA and Min. Temp. MA) were the least important
291 (Figure 5). The decrease in variable importance (from sex to distance to Min. Temp. MA) was
292 gradual.

293

294 [Figure 5 here]

295

296 4 DISCUSSION

297 Using the conditional random forests model, we found that elephant sex and the presence of
298 roads or human settlements were the strongest predictors of the length and speed of elephant
299 movements. Contrary to the main hypothesis, the tagged elephants did not clearly undertake any
300 seasonal migration. Our model confirms that elephants in PNOK experience human disturbances.
301 Consequently, despite taking into account the importance that touristic activities have on habitat
302 conservation, regulating and restricting them—in time and space—to avoid human-elephant
303 conflicts and disturbance could notably contribute to elephant conservation. We acknowledge that a
304 larger sample size, both in duration of the study and in the number of tagged individuals is desirable
305 to understand the ecology of the population of elephants in OKNP. Therefore, these results should be
306 interpreted with caution as representing these six individuals and not the entire population.
307 Nevertheless, although the r^2 of 0.61 suggests that other unmeasured variables also influence
308 elephant movements, this value is similar to other studies employing a restricted number of
309 individuals and input variables (Bohrer et al., 2014).

310 Female elephants undertook longer daytime movements than males contradicting results of
311 previous studies (Kolowski et al., 2010; Mills et al., 2018), which we hypothesize may be first due to
312 our low sample size for males, and secondly due to the possible reproductive status for some females
313 and the resources they are accessing during the study period. The ecological strategies of lactating
314 females are generally focused on the intake of sufficient water and calories to maximise milk
315 production (for data on elephants' age estimates see Appendix A in supplementary materials). In
316 addition, the high availability of water (de Beer & van Aarde, 2008) and the sexual dimorphism of
317 the species, would allow males to thrive in habitats of low water availability, which also explain the

318 relatively shorter distances travelled by our males. In any case, these differences between sexes
319 should be interpreted with caution due to our particular small sample size for males.

320 With regard to the distance to roads, the partial dependence plots indicated that forest elephants
321 did not select the vicinity of roads for resting, feeding or social purposes as has been observed in
322 areas with controlled hunting (Kolowski et al., 2010). The use of the dirt road —the only path that
323 supports road traffic— is restricted to the OKNP and lodge staff and to the few tourists visiting the
324 lodge. Consequently, traffic along the road is sporadic and occurs during the daytime. Nevertheless,
325 according to the sequence of fixes obtained during the study period, our collared elephants did not
326 use this road for travelling, because most of movements were perpendicular to the road. Protected
327 roads in OKNP are not a barrier for elephants' movement, as opposed to those found outside other
328 protected areas (Blake et al., 2008). However, our results indicate that even the necessary activities
329 performed in the park, such as eco-monitoring and anti-poaching patrolling, and the visits of small
330 groups of tourists, affects the speed of elephant movements to some extent.

331 Faster displacements occurred near the lodge complex and, neglecting the long displacements of
332 female 4 (F4) (see Figure 1), in areas between 22 and 31 km from them. It has been suggested that
333 human activities alter elephant behaviour for short time spans (Wrege, Rowland, Thompson, &
334 Batruch, 2010), emphasizing the possibility that park activities may cause elephants to change their
335 behaviour. However, the lodge is located near some of the biggest baobabs in the park, which in turn
336 may favour faster movements as elephants travel to habitats with abundant resources and/or for
337 social interactions. Despite the uncertainty, near these areas it would be recommendable to perform
338 tourist activities with special care to avoid frequent disruption to the forest elephants' normal
339 behaviour.

340 Although most of the GPS locations were obtained in the same range of elevations (with the
341 exception of female 4 – F4), below 350 m a.s.l. there was a sharp decrease in the speed of the

342 elephants. This fact, combined with the information described in the partial dependence plots for the
343 meteorological variables, suggests that water presence in low elevation flood plains may slow down
344 elephant movements but also because they positively select these habitats to maintain social
345 relationships (Fishlock, 2010; Fishlock & Lee, 2013) and to satisfy thermoregulatory needs
346 (Kuwong, 2014). On the contrary, most of the displacements were made following the SE-NW axis
347 of the Mambili River basin, where the edge of the riverine area facilitates their movement.
348 Apparently, such movement patterns coincide with those described for a number of mammals that
349 employ riparian corridors as ‘highways’ during their displacements (Sánchez-Montoya et al.2016), a
350 behaviour that has also been occasionally reported for African elephants. These two phenomena
351 related to flooded plains and movement through riverine habitats do not necessarily have to be
352 contemporary or to occur at the same spatial scale. Notwithstanding, forest elephant at the OKNP did
353 not perform migrations in the same way as savannah elephants do (Birkett et al., 2012; Bohrer et al.,
354 2014), although they are still affected by weather patterns. The moving average of the maximum
355 temperature (Max. Temp. MA) affected elephant linear displacement above 31.5 °C resulting in an
356 abrupt decrease in distance moved between fixes the linear displacement, which might be explained
357 by their low surface-to-volume ratio that affects heat dissipation (Williams, 1990). Although with a
358 less pronounced pattern, the temperature and the minimum temperature (Temp. MA and Min. Temp.
359 MA) were positively related to activity and distance moved. Finally, precipitation was inversely
360 related to elephant movement; during the periods of highest rainfall, the elephants moved more
361 slowly. We hypothesise that, aside from flooding, this pattern would highlight that elephants are
362 more inclined to rest during storms, or to seek cover and hide if they suffer major disturbances (i.e.
363 poaching).

364 The study of GPS collared elephants has shed some light on the drivers of linear displacements
365 of the forest elephant in the OKNP in regard to a suite of geographical, meteorological and
366 anthropogenic variables. Sex, land cover and three anthropogenic, two geographic and four

367 meteorological variables described their daily linear displacements. Despite our low sample size, the
368 present study produced some interesting results. Previous studies found that unprotected roads have a
369 barrier effect in elephant displacements, but not in the case of roads located inside protected areas
370 (Blake et al., 2008; Kolowski et al., 2010; Poulsen et al., 2018; Schuttler et al., 2012). Given that our
371 road is a protected road inside a National Park our results are consistent with these previous studies.
372 Nevertheless, the partial dependence plots exposed a disruptive effect of its presence because our
373 elephants moved faster near the roads. Our elephants showed a trend to move slower in savannahs
374 than in any other habitat what perfectly agrees with what Mills et al. (2018) found at Wonga Wongué
375 Presidential Reserve in Gabon. Contrary to what was found in the studies cited above, we did not
376 find a strict seasonal movement pattern, characterised by sustained long unidirectional
377 displacements, apart from the differences found for temperature and rainfall. We consider our results
378 to be of great interest for wildlife management plans because they highlight drivers of daily
379 movement distance for the forest elephant, which is especially relevant because daily displacements
380 performed by elephants would be one of the factors affecting their vulnerability to poachers
381 (Goldenberg et al., 2018). We found evidences that the activities performed during the standard
382 operation of the park and the distribution of the different habitat patches affected the daily linear
383 displacement of the collared elephants. Notwithstanding, the notorious importance of tourism related
384 employment in remote rural areas (reducing poverty, improving the social welfare of local
385 communities and promoting biodiversity conservation (Dinets & Hall, 2018; Snyman, 2012), the
386 strict control of the tourist activity schedules and the application of especial care when performing
387 these activities may prevent unnecessary disturbances to vulnerable fauna. Consequently, human
388 activities and displacements within the park should be limited—in time and space—as much as
389 possible, this would be especially important for tourist groups, by avoiding large groups of people
390 and trying to concentrate their activities at specific places for given periods of time. Furthermore,
391 proactive vegetation management, such as the ongoing controlled fires for savannahs' maintenance

392 or the recovering of degraded patches could be used to encourage different activities. Overall, we
393 expect these results to favour grounded decisions in the process of management of this endangered
394 species.

395

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404

405 6 REFERENCES

- 406 Arlot, S., & Celisse, A. (2010). A survey of cross-validation procedures for model selection.
407 *Statistics Surveys*, 4, 40–79. <https://doi.org/10.1214/09-SS054>
- 408 Aveling, C., & Froment, J. M. (2001). L'extension du Parc National d'Odzala, une opportunité de
409 développement. *Canopée*, 20, 13–14. JOUR.
- 410 Bermejo, M. (1999). Status and conservation of primates in Odzala National Park, Republic of the
411 Congo. *Oryx*, 33(4), 324–332. JOUR.
- 412 Birkett, P., Vanak, A., Muggeo, V., Ferreira, S., & Slotow, R. (2012). Animal Perception of Seasonal
413 Thresholds: Changes in Elephant Movement in Relation to Rainfall Patterns. *PLoS ONE*, 7(6),
414 e38363. <https://doi.org/10.1371/journal.pone.0038363>

- 415 Blake, S. (2002). The ecology of forest elephant distribution and its implications for conservation.
416 *University of Edinburgh, Edinburgh, Scotland*, 319.
417 <https://doi.org/10.1017/S0022112097006745>
- 418 Blake, S., Deem, S. L., Strindberg, S., Maisels, F., Momont, L., Isia, I. B., ... Kock, M. D. (2008).
419 Roadless wilderness area determines forest elephant movements in the Congo Basin. *PLoS*
420 *ONE*, 3(10). <https://doi.org/10.1371/journal.pone.0003546>
- 421 Blake, S., Douglas-Hamilton, I., & Karesh, W. B. (2001). GPS telemetry of forest elephants in
422 Central Africa: Results of a preliminary study. *African Journal of Ecology*, 39(2), 178–186.
423 <https://doi.org/10.1046/j.1365-2028.2001.00296.x>
- 424 Blake, S., Strindberg, S., Boudjan, P., Makombo, C., Bila-Isia, I., Ilambu, O., ... Maisels, F. (2007).
425 Forest elephant crisis in the Congo Basin. *PLoS Biology*, 5(4), 945–953.
426 <https://doi.org/10.1371/journal.pbio.0050111>
- 427 Bohrer, G., Beck, P. S. P., Ngene, S. M. S., Skidmore, A. K. A., & Douglas-Hamilton, I. (2014).
428 Elephant movement closely tracks precipitation-driven vegetation dynamics in a Kenyan forest-
429 savanna landscape. *Movement Ecology*, 2(1), 2. <https://doi.org/10.1186/2051-3933-2-2>
- 430 Boukoulou, H., Mbete, P., Mbete, R., Ngokaka, C., Akouango, F., Excelh, B. K. R., & Voudibio, J.
431 (2012). Conflit Homme/Éléphant: étude de cas dans le village Miélékouka au Nord du Parc
432 National d'Odzala Kokoua (Congo). *Journal of Applied Biosciences*, 50, 3478–3484.
- 433 Breiman, L. (2001). Random forests. *Machine Learning*, 45(1), 5–32.
434 <https://doi.org/http://dx.doi.org/10.1023/A:1010933404324>
- 435 Breuer, T., Maisels, F., & Fishlock, V. (2016). The consequences of poaching and anthropogenic
436 change for forest elephants. *Conservation Biology*, 30(5), 1019–1026.
437 <https://doi.org/10.1111/cobi.12679>

438 Buij, R., McShea, W. J., Campbell, P., Lee, M. E., Dallmeier, F., Guimondou, S., ... Alonso, A.
439 (2007). Patch-occupancy models indicate human activity as major determinant of forest
440 elephant *Loxodonta cyclotis* seasonal distribution in an industrial corridor in Gabon. *Biological*
441 *Conservation*, 135(2), 189–201. <https://doi.org/10.1016/j.biocon.2006.10.028>

442 CITES. (2012). Convention on International Trade in Endangered Species of Wild Fauna and Flora,
443 3 March 1973. In *Documents in International Environmental Law* (Vol. 6, pp. 661–680).
444 <https://doi.org/10.1017/cbo9781139171380.035>

445 CITES. (2016). *Convention on International Trade in Endangered Species of Wild Fauna and Flora*
446 (Vol. 6).

447 Clark, C. J., Poulsen, J. R., Malonga, R., & Elkan, P. W. (2009). Logging concessions can extend the
448 conservation estate for central African tropical forests. *Conservation Biology*, 23(5), 1281–
449 1293. <https://doi.org/10.1111/j.1523-1739.2009.01243.x>

450 Crooks, K. R., Burdett, C. L., King, S. R. B., Di Marco, M., Boitani, L., Theobald, D. M., &
451 Rondinini, C. (2017). Quantification of habitat fragmentation reveals extinction risk in
452 terrestrial mammals. *Proceedings of the National Academy of Sciences*, 114(29), 7635–7640.
453 <https://doi.org/10.1073/pnas.1705769114>

454 D'Eon, R., Serrouya, R., Smith, G., & Kochanny, C. (2002). GPS Radiotelemetry Error and Bias in
455 Mountainous Terrain. *Wildlife Society Bulletin*, 30(2), 430–439.

456 de Beer, Y., & van Aarde, R. J. (2008). Do landscape heterogeneity and water distribution explain
457 aspects of elephant home range in southern Africa's arid savannas? *Journal of Arid*
458 *Environments*, 72(11), 2017–2025. <https://doi.org/10.1016/j.jaridenv.2008.07.002>

459 de Knegt, H. J., van Langevelde, F., Skidmore, A. K., Delsink, A., Slotow, R., Henley, S., ... Prins,
460 H. H. T. (2011). The spatial scaling of habitat selection by African elephants. *Journal of Animal*
461 *Ecology*, 80(1), 270–281. <https://doi.org/10.1111/j.1365-2656.2010.01764.x>

- 462 Di Marco, M., Buchanan, G. M., Szantoi, Z., Holmgren, M., Grottolo Marasini, G., Gross, D., ...
463 Rondinini, C. (2014). Drivers of extinction risk in African mammals: the interplay of
464 distribution state, human pressure, conservation response and species biology. *Philosophical*
465 *Transactions of the Royal Society of London B: Biological Sciences*, 369(1643).
- 466 Dinets, V., & Hall, J. (2018). Mammalwatching: A new source of support for science and
467 conservation. *International Journal of Biodiversity and Conservation*, 10(4), 154–160.
468 <https://doi.org/10.5897/ijbc2017.1162>
- 469 Elliot, N. B., Cushman, S. A., Loveridge, A. J., Mtare, G., & Macdonald, D. W. (2014). Movements
470 vary according to dispersal stage, group size, and rainfall: the case of the African lion. *Ecology*,
471 95(10), 2860–2869. <https://doi.org/10.1890/13-1793.1>
- 472 Epskamp, S., Cramer, A. O. J., Waldorp, L. J., Schmittmann, V. D., & Borsboom, D. (2012). qgraph:
473 Network Visualizations of Relationships in Psychometric Data. *Journal of Statistical Software*,
474 48(4), 1–18. Retrieved from <http://www.jstatsoft.org/v48/i04/>
- 475 Fishlock, V. L. (2010). Bai Use in Forest Elephants (*Loxodonta africana cyclotis*): A Thesis
476 Submitted for the Degree of Doctor of Philosophy, (August).
- 477 Fishlock, V., & Lee, P. C. (2013). Forest elephants: Fission-fusion and social arenas. *Animal*
478 *Behaviour*, 85(2), 357–363. <https://doi.org/10.1016/j.anbehav.2012.11.004>
- 479 Fox, J. (2010). polycor: Polychoric and Polyserial Correlations. Retrieved from [http://cran.r-](http://cran.r-project.org/package=polycor)
480 [project.org/package=polycor](http://cran.r-project.org/package=polycor)
- 481 Friedman, J. H. (2001). Greedy function approximation: A gradient boosting machine. *Annals of*
482 *Statistics*, 29(5), 1189–1232. <https://doi.org/10.1214/aos/1013203451>
- 483 Fruchterman, T. M. J., & Reingold, E. M. (1991). Graph drawing by force-directed placement.
484 *Software: Practice and Experience*, 21(11), 1129–1164.

485 <https://doi.org/10.1002/spe.4380211102>

486 Gobush, K. S., Mutayoba, B. M., & Wasser, S. K. (2008). Long-term impacts of poaching on
487 relatedness, stress physiology, and reproductive output of adult female African elephants.
488 *Conservation Biology*, 22(6). <https://doi.org/10.1111/j.1523-1739.2008.01035.x>

489 Goldenberg, S. Z., Douglas-Hamilton, I., Daballen, D., & Wittemyer, G. (2016). Challenges of using
490 behavior to monitor anthropogenic impacts on wildlife: a case study on illegal killing of African
491 elephants. *Animal Conservation*. <https://doi.org/10.1111/acv.12309>

492 Goldenberg, S. Z., Douglas-Hamilton, I., & Wittemyer, G. (2018). Inter-generational change in
493 african elephant range use is associated with poaching risk, primary productivity and adult
494 mortality. *Proceedings of the Royal Society B: Biological Sciences*, 285(1879).
495 <https://doi.org/10.1098/rspb.2018.0286>

496 Gonzalez-Voyer, A., González-Suárez, M., Vilà, C., & Revilla, E. (2016). Larger brain size
497 indirectly increases vulnerability to extinction in mammals. *Evolution; International Journal of*
498 *Organic Evolution*, 70(6), 1364–1375. <https://doi.org/10.1111/evo.12943>

499 Graham, M. D., Douglas-Hamilton, I., Adams, W. M., & Lee, P. C. (2009). The movement of
500 African elephants in a human-dominated land-use mosaic. *Animal Conservation*, 12(5), 445–
501 455. <https://doi.org/10.1111/j.1469-1795.2009.00272.x>

502 Harris, G., Thirgood, S., Hopcraft, J. G. C., Cromsigt, J. P. G. M., & Berger, J. (2009). Global
503 decline in aggregated migrations of large terrestrial mammals. *Endangered Species Research*,
504 7(1), 55–76. <https://doi.org/10.3354/esr00173>

505 Hothorn, T., Hornik, K., & Zeileis, A. (2006). Unbiased recursive partitioning: A conditional
506 inference framework. *Journal of Computational and Graphical Statistics*, 15(3), 651–674.
507 <https://doi.org/10.1198/106186006X133933>

- 508 Johnson, D. D. P., Kays, R., Blackwell, P. G., & Macdonald, D. W. (2002). Does the resource
509 dispersion hypothesis explain group living? *Trends in Ecology and Evolution*, *17*(12), 563–570.
510 [https://doi.org/10.1016/S0169-5347\(02\)02619-8](https://doi.org/10.1016/S0169-5347(02)02619-8)
- 511 Kolowski, J. M., Blake, S., Kock, M. D., Lee, M. E., Henderson, A., Honorez, A., & Alonso, A.
512 (2010). Movements of four forest elephants in an oil concession in Gabon, Central Africa.
513 *African Journal of Ecology*, *48*(4), 1134–1138. [https://doi.org/10.1111/j.1365-](https://doi.org/10.1111/j.1365-2028.2009.01204.x)
514 [2028.2009.01204.x](https://doi.org/10.1111/j.1365-2028.2009.01204.x)
- 515 Kuwong, M. V. (2014). Microclimate mosaic and its influence on behaviour of free-living African
516 forest elephants (*Loxodonta africana cyclotis*). *PhD Thesis*, 149.
- 517 Laurance, W. F., Croes, B. M., Tchignoumba, L., Lahm, S. A., Alonso, A., Lee, M. E., ... Ondzeano,
518 C. (2006). Impacts of roads and hunting on central African rainforest mammals. *Conservation*
519 *Biology*, *20*(4), 1251–1261. <https://doi.org/10.1111/j.1523-1739.2006.00420.x>
- 520 Liaw, A., & Wiener, M. (2002). Classification and regression by randomForest. *R News*, *3*(2), 18–22.
- 521 Loarie, S. R., Aarde, R. J. Van, & Pimm, S. L. (2009). Fences and artificial water affect African
522 savannah elephant movement patterns. *Biological Conservation*, *142*(12), 3086–3098.
523 <https://doi.org/10.1016/j.biocon.2009.08.008>
- 524 Maisels, F., Strindberg, S., Blake, S., Wittemyer, G., Hart, J., Williamson, E. A., ... Warren, Y.
525 (2013). Devastating Decline of Forest Elephants in Central Africa. *PLoS ONE*, *8*(3).
526 <https://doi.org/10.1371/journal.pone.0059469>
- 527 May, R., Dandy, G., & Maier, H. (2011). Review of Input Variable Selection Methods for Artificial
528 Neural Networks. In K. Suzuki (Ed.), *Artificial Neural Networks - Methodological Advances*
529 *and Biomedical Applications*. (p. 362). InTech. <https://doi.org/http://dx.doi.org/10.5772/644>
- 530 Mbete, P., Ngokaka, C., Ntounta, F. A., & Vouidibio, J. (2010). Evaluation des quantités de gibiers

531 prélevées autour du Parc National d ' Odzala-Kokoua et leurs impacts sur la. *Journal of Animal*
532 *& Plant Sciences*, 8(3), 1061–1069.

533 Metsio Sienne, J., Buchwald, R., & Wittemyer, G. (2014). Differentiation in mineral constituents in
534 elephant selected versus unselected water and soil resources at Central African bais (forest
535 clearings). *European Journal of Wildlife Research*, 60(2), 377–382.
536 <https://doi.org/10.1007/s10344-013-0781-0>

537 Milliken, T., Burn, R. W., Underwood, F. M., & Sangalakula, L. (2013). ETIS report of TRAFFIC -
538 Monitoring Illegal Trade in Ivory. *Cites*, 2, 1–30.

539 Mills, E. C., Poulsen, J. R., Michael Fay, J., Morkel, P., Clark, C. J., Meier, A., ... White, L. J. T.
540 (2018). Forest elephant movement and habitat use in a tropical forest-grassland mosaic in
541 Gabon. *PLoS ONE*, 13(7), 1–17. <https://doi.org/10.1371/journal.pone.0199387>

542 Muñoz-Mas, R., Fukuda, S., Pórtolés, J., & Martínez-Capel, F. (2018). Revisiting probabilistic
543 neural networks: a comparative study with support vector machines and the microhabitat
544 suitability for the Eastern Iberian chub (*Squalius valentinus*). *Ecological Informatics*, 43, 24–37.
545 <https://doi.org/10.1016/J.ECOINF.2017.10.008>

546 Muñoz-Mas, R., Fukuda, S., Vezza, P., & Martínez-Capel, F. (2016). Comparing four methods for
547 decision-tree induction: A case study on the invasive Iberian gudgeon (*Gobio lozanoi*; Doadrio
548 and Madeira, 2004). *Ecological Informatics*, 34. <https://doi.org/10.1016/j.ecoinf.2016.04.011>

549 Poulsen, J. R., Koerner, S. E., Moore, S., Medjibe, V. P., Blake, S., Clark, C. J., ... White, L. J. T.
550 (2017). Poaching empties critical Central African wilderness of forest elephants. *Current*
551 *Biology*, 27(4), R134–R135. <https://doi.org/10.1016/j.cub.2017.01.023>

552 Poulsen, J. R., Rosin, C., Meier, A., Mills, E., Nuñez, C. L., Koerner, S. E., ... Sowers, M. (2018).
553 Ecological consequences of forest elephant declines for Afrotropical forests. *Conservation*
554 *Biology*, 32(3), 559–567. <https://doi.org/10.1111/cobi.13035>

555 R Core Team. (2015). R: A language and environment for statistical computing. Vienna, Austria.: R
556 Foundation for Statistical Computing. Retrieved from <http://www.r-project.org/>

557 Ripple, W. J., Abernethy, K., Betts, M. G., Chapron, G., Dirzo, R., Galetti, M., ... Young, H. (2016).
558 Bushmeat hunting and extinction risk to the world's mammals. *Royal Society Open Science*,
559 3(10). Retrieved from <http://rsos.royalsocietypublishing.org/content/3/10/160498.abstract>

560 Sánchez-Montoya, M. M., Moleón, M., Sánchez-Zapata, J. A., & Tockner, K. (2016). Dry riverbeds:
561 corridors for terrestrial vertebrates. *Ecosphere*, 7(10), e01508. <https://doi.org/10.1002/ecs2.1508>

562 Schuttler, S. G., Blake, S., & Eggert, L. S. (2012). Movement Patterns and Spatial Relationships
563 Among African Forest Elephants, 44(4), 445–448. [https://doi.org/10.1111/j.1744-](https://doi.org/10.1111/j.1744-7429.2012.00889.x)
564 [7429.2012.00889.x](https://doi.org/10.1111/j.1744-7429.2012.00889.x)

565 Shannon, G., Page, B. R., Duffy, K. J., & Slotow, R. (2006). The role of foraging behaviour in the
566 sexual segregation of the African elephant. *Oecologia*, 150(2), 344–354.
567 <https://doi.org/10.1007/s00442-006-0521-1>

568 Short, J. C. (1983). Density and seasonal movements of forest elephant (*Loxodonta africana cyclotis*,
569 Matschie) in Bia National Park, Ghana. *African Journal of Ecology*, 21(3), 175–184.
570 <https://doi.org/10.1111/j.1365-2028.1983.tb01179.x>

571 Snyman, S. L. (2012). The role of tourism employment in poverty reduction and community
572 perceptions of conservation and tourism in southern Africa. *Journal of Sustainable Tourism*,
573 20(3), 395–416. <https://doi.org/10.1080/09669582.2012.657202>

574 Stokes, E. J., Strindberg, S., Bakabana, P. C., Elkan, P. W., Iyenguet, F. C., Madzoké, B., ... Rainey,
575 H. J. (2010). Monitoring great ape and elephant abundance at large spatial scales: measuring
576 effectiveness of a conservation landscape. *PloS One*, 5(4), e10294.
577 <https://doi.org/10.1371/journal.pone.0010294>

- 578 Strobl, C., Boulesteix, A.-L., Zeileis, A., & Hothorn, T. (2007). Bias in random forest variable
579 importance measures: illustrations, sources and a solution. *BMC Bioinformatics*, 8, 25.
580 <https://doi.org/10.1186/1471-2105-8-25>
- 581 Strobl, C., Hothorn, & Zeileis, A. (2009). Party on! *R Journal*, 1(2), 14–17. Retrieved from internal-
582 pdf://party
- 583 Turkalo, A. K. (2013). Estimating forest elephant age. *African Journal of Ecology*, 51(3), 501–505.
584 <https://doi.org/10.1111/aje.12087>
- 585 Turkalo, A. K., Wrege, P. H., & Wittemyer, G. (2013). Long-Term monitoring of dzanga bai forest
586 elephants: Forest clearing use patterns. *PLoS ONE*, 8(12), 1–12.
587 <https://doi.org/10.1371/journal.pone.0085154>
- 588 Wasser, S. K., Brown, L., Mailand, C., Mondol, S., Clark, W., Laurie, C., & Weir, B. S. (2015).
589 Genetic assignment of large seizures of elephant ivory reveals Africa’s major poaching
590 hotspots. *Science*, 349(6243), 84–87. <https://doi.org/10.1126/science.aaa2457>.Genetic
- 591 White, L. J.-T. (1992). *Vegetation history and logging disturbance: effects on rain forest in the Lope*
592 *Reserve, Gabon (with special emphasis on elephants and apes)*.
- 593 Williams, T. M. (1990). Heat transfer in elephants: thermal partitioning based on skin temperature
594 profiles. *Journal of Zoology*, 222(2). <https://doi.org/10.1111/j.1469-7998.1990.tb05674.x>
- 595 Wittemyer, G., Northrup, J. M., Blanc, J., Douglas-Hamilton, I., Omondi, P., & Burnham, K. P.
596 (2014). Illegal killing for ivory drives global decline in African elephants. *Proceedings of the*
597 *National Academy of Sciences*, 111(36), 13117–13121.
598 <https://doi.org/10.1073/pnas.1403984111>
- 599 Wrege, P. H., Rowland, E. D., Thompson, B. G., & Batruch, N. (2010). Use of acoustic tools to
600 reveal otherwise cryptic responses of forest elephants to oil exploration. *Conservation Biology*,

601 24(6). <https://doi.org/10.1111/j.1523-1739.2010.01559.x>

602 Young, K. D., Ferreira, S. M., & van Aarde, R. J. (2009). Elephant spatial use in wet and dry
603 savannas of southern Africa. *Journal of Zoology*, 278(3), 189–205.
604 <https://doi.org/10.1111/j.1469-7998.2009.00568.x>

605

606 **TABLES**

607

608 Table 1. Summary and units of input variables employed in the analysis of the movement patterns
 609 with conditional random forests. MA corresponds to moving average, Min. to minimum, Qu. to

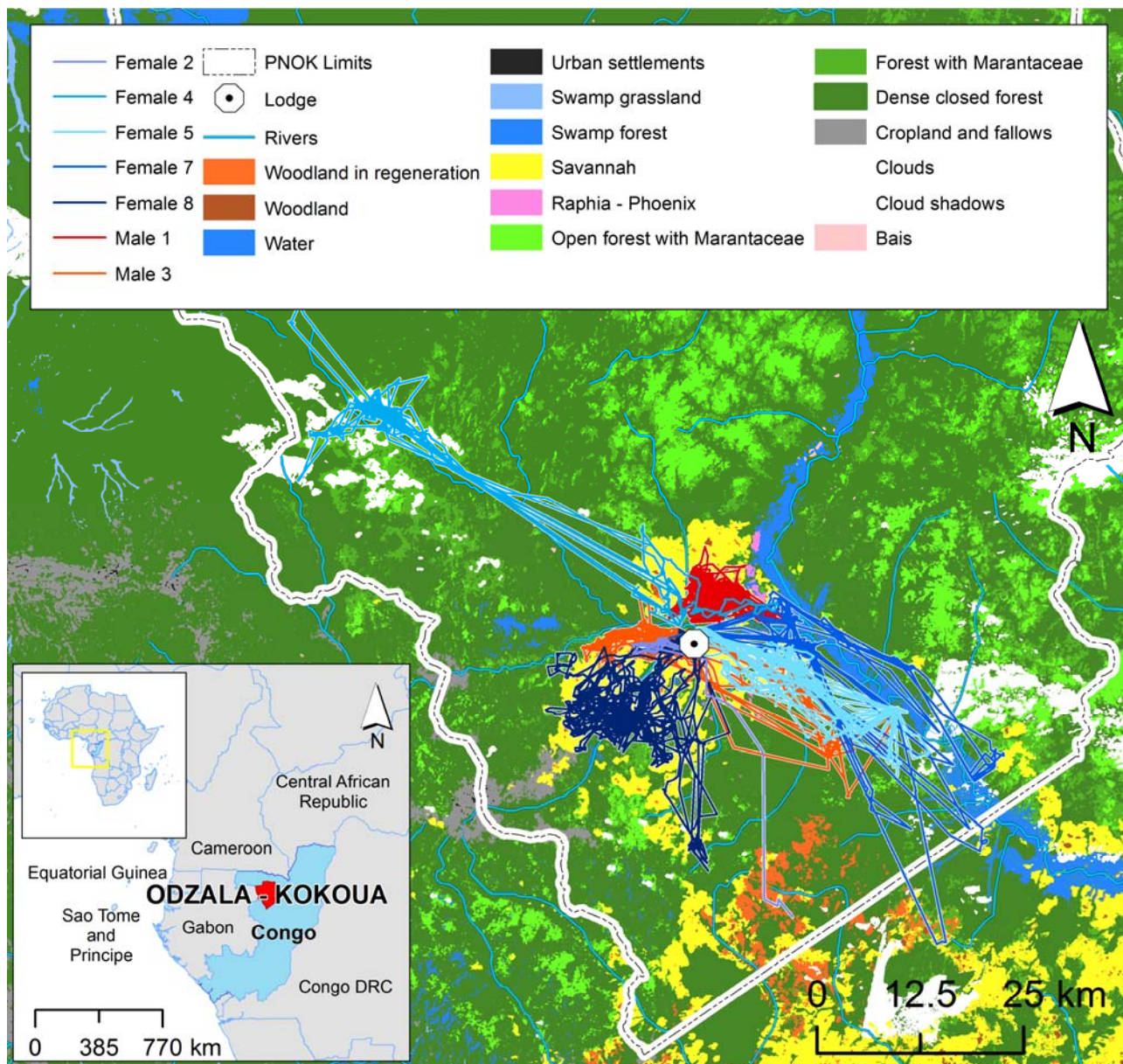
610

Quartile and Max. to maximum.

Variable	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
Linear displacement (km/24 h)	0.00	0.81	1.40	1.88	2.40	15.41
Temp. (°C)	21.67	25.12	26.00	26.07	27.14	30.44
Max. Temp. (°C)	23.31	29.49	31.21	30.87	32.39	35.71
Min. Temp. (°C)	13.08	20.28	20.97	20.86	21.56	26.94
Precipitation (mm)	0.00	0.00	0.37	4.07	4.03	61.98
Temp. MA (°C)	25.15	25.48	25.82	26.03	26.56	27.16
Max. Temp. MA (°C)	28.60	30.30	30.98	30.88	31.64	32.25
Min. Temp. MA (°C)	17.90	20.51	20.96	20.80	21.24	22.32
Precipitation MA (mm)	0.26	2.17	4.33	4.09	5.65	8.46
Elevation (m a.s.l.)	352.00	377.00	393.00	420.10	445.00	740.00
Slope (°)	0.00	1.08	2.21	2.40	3.58	13.91
Distance to rivers (km)	0.00	1.03	1.59	1.83	2.44	7.48
Distance to villages (km)	8.97	22.79	31.52	28.56	34.18	42.35
Distance to roads (km)	0.03	4.08	7.42	10.73	14.68	41.16
Distance to lodge (km)	0.20	6.32	11.12	14.61	17.38	63.00
Season	Short dry, short rainy, long dry & long rainy					
Land cover type	Bais, dense closed forest, forest with Marantaceae, open forest with Marantaceae, Raphia – Phoenix, savannah, swamp forest, water, woodland, woodland in regeneration + urban settlements, croplands and fallows, swamp grassland (never visited)					
Sex	Male & female					

611

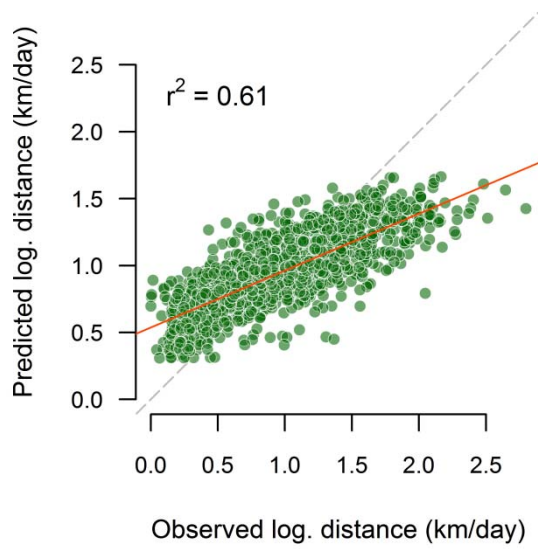
612 **FIGURES:**



613

614 Figure 1. Location of the Odzala-Kokoua National Park (OKNP) within the African continent and
 615 general view of the elephant trajectories, land cover, rivers and villages in the environs of the OKNP.

616



617

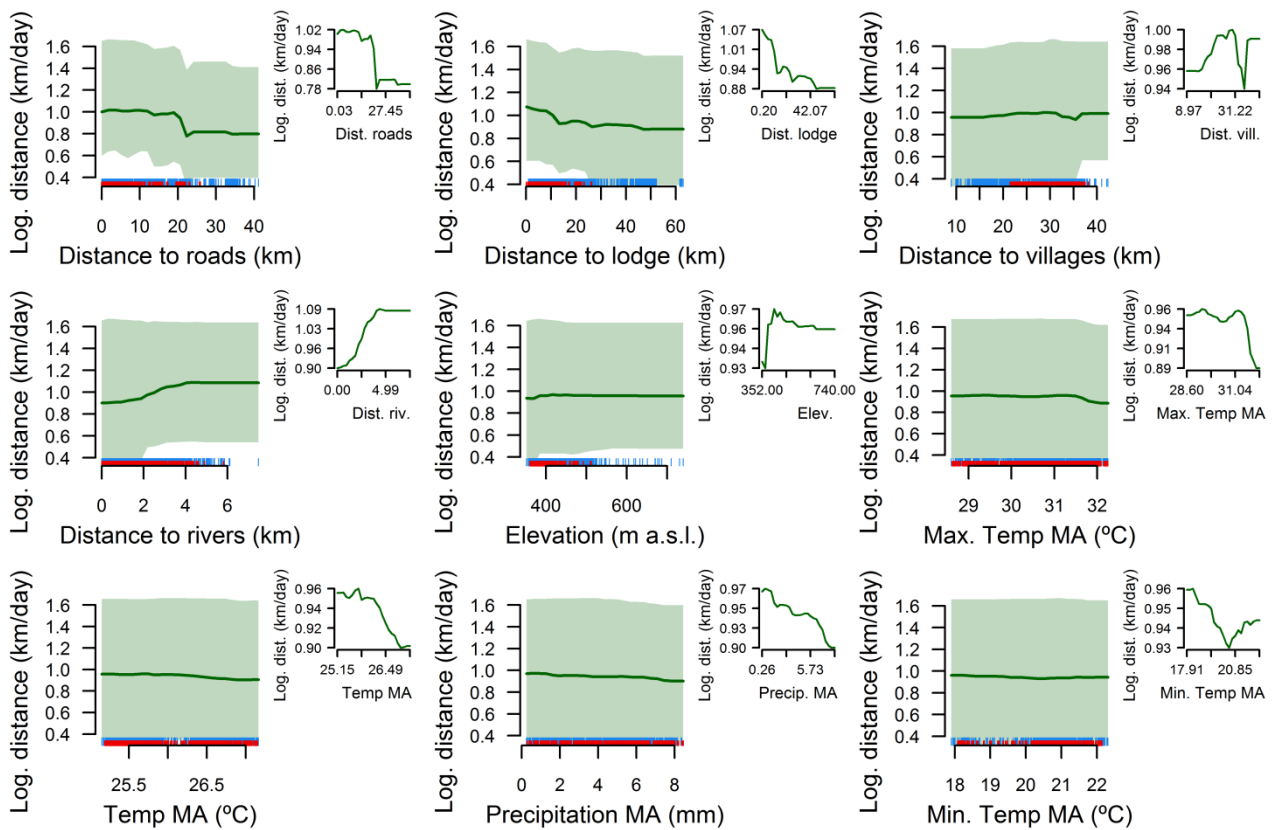
618

Figure 2. Observed versus predicted mean daily linear displacement by the forest elephants and

619

coefficient of determination (r^2). The red line: obtained data, grey line: reference $r = 1$.

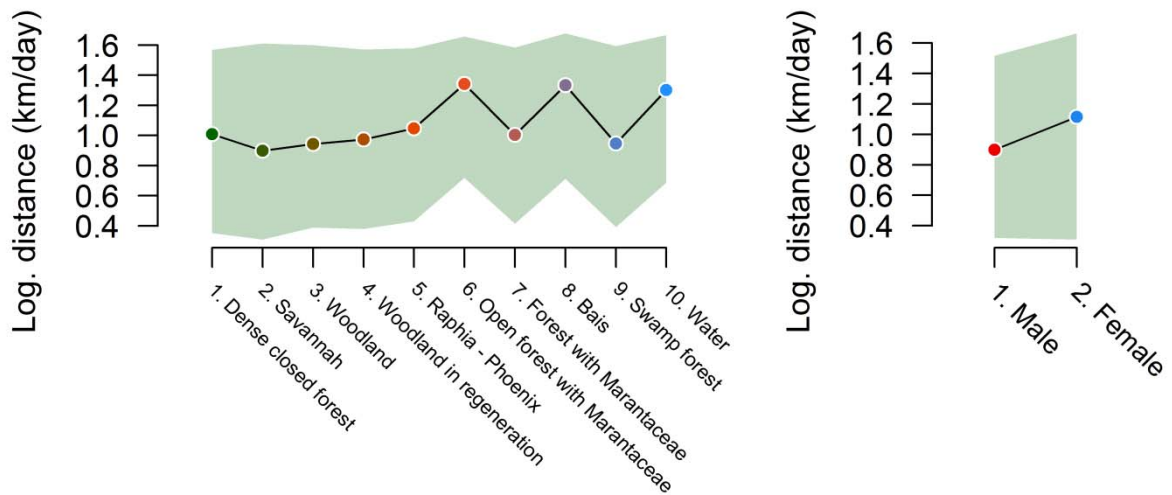
620



621

622 Figure 3. Partial dependence plots for the most relevant continuous variables selected during
 623 conditional random forests optimisation. MA corresponds to moving average, Min. to minimum and
 624 Max. to maximum. Tick marks represent data contained in the training dataset. Blue corresponds to
 625 females and red to males. The shaded area depicts the range of the predictions obtained for each
 626 permuted value. In order to clarify the variable effect, the mean trend (central line in the larger plots)
 627 is depicted next to main plots.

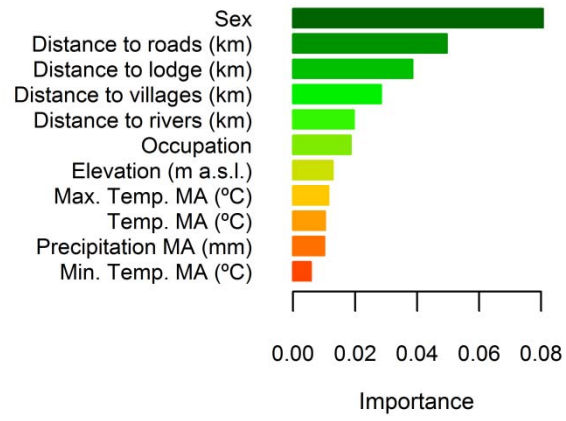
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629

630 Figure 4. Partial dependence plots for the most relevant categorical variables selected during
 631 conditional random forests optimisation. The shaded area depicts the range of the predictions
 632 obtained for each permuted value. Colour code differentiates each category.

633



634

635

Figure 5. Variable importance obtained with conditional random forests.

636

1 **Appendices**

2 APPENDIX A – INDIVIDUALS DATA SUMMARY TABLE

3 Biological characteristics (age and sex) and main tracking data obtained per individual (data
4 of the first and last fixes gathered, total number of fixes obtained and total number of days tracked)
5 are shown in table A1.

6

7 [Table A1 Here]

8

9 APPENDIX B – CORRELATION AMONG ENVIRONMENTAL COVARIATES

10 The correlation among environmental variables has been characterised with a force-directed
11 graph (Fruchterman & Reingold, 1991) (Figure A1). This graph depicts the correlation between the
12 geographical, meteorological and anthropogenic variables, which revealed the presence of two
13 groups of variables encompassing, the geographical and anthropogenic variables on the one hand,
14 and the meteorological variables on the other hand. Relevant correlations were obtained for
15 temperature and maximum temperature and for season and the moving average of the maximum
16 temperature ($r^2 = 0.83$ and $r^2 = 0.82$). The tracked males remained near the Lango Lodge (see Figure
17 1 in the main text). Therefore, sex and distance to lodge presented a relevant correlation ($|r^2| = 0.82$).
18 The latter also correlated with distance to road ($r^2 = 0.90$). Correlations were obtained with the *R*
19 package *polycor* (Fox, 2010), which is specially designed to handle continuous and categorical data
20 and the force-directed graph was developed using the functions included in the *R* package *qgraph*
21 (Epskamp, Cramer, Waldorp, Schmittmann, & Borsboom, 2012).

22

23 [Figure B1 Here]

24

25 APPENDIX C – LINEAR DISPLACEMENT PER ELEPHANT

26 The mean distance performed by the tracked elephants was 1.87 km per day. Female displacements
27 were longer (2.05 ± 1.82), although they presented higher variability than that of males (1.65 ± 1.25).
28 The smallest displacement for both sexes was close to zero. Female 4 performed the longest
29 displacement per day, which correspond to the three outliers depicted in Figure C1. Additional
30 details about these displacements can be found in the main text.

31

32 [Figure C1 Here]

33

34 REFERENCES

- 35 Epskamp, S., Cramer, A. O. J., Waldorp, L. J., Schmittmann, V. D., & Borsboom, D. (2012). qgraph:
36 Network Visualizations of Relationships in Psychometric Data. *Journal of Statistical Software*,
37 48(4), 1–18. Retrieved from <http://www.jstatsoft.org/v48/i04/>
- 38 Fox, J. (2010). polycor: Polychoric and Polyserial Correlations. Retrieved from [http://cran.r-](http://cran.r-project.org/package=polycor)
39 [project.org/package=polycor](http://cran.r-project.org/package=polycor)
- 40 Fruchterman, T. M. J., & Reingold, E. M. (1991). Graph drawing by force-directed placement.
41 *Software: Practice and Experience*, 21(11), 1129–1164.
42 <https://doi.org/10.1002/spe.4380211102>

43

44 **APPENDIX TABLES AND FIGURES:**

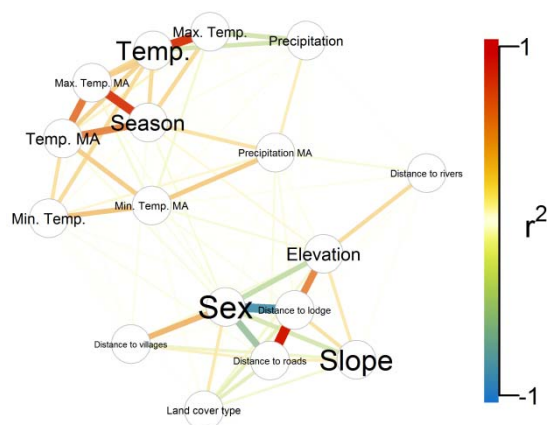
45

46 Table A 1. Sex, estimated age in years, trapping and collar-fitting data (Start), last fix data (End),
 47 total number of fixes obtained (NFixes) and total number of days tracked (NDays) per individual.

48

Individual	Sex	Age	Start	End	NFixes	NDays
M1	Male	30-35	20/11/2014	11/3/2016	5685	477
F2	Female	10-15	21/11/2014	11/3/2016	112	476
M3	Male	20-25	22/11/2014	14/3/2016	2597	478
F4	Female	40+	25/11/2014	28/3/2016	1155	489
F5	Female	10-12	28/11/2014	11/3/2016	1439	469
F7	Female	10-15	29/11/2014	1/4/2016	1166	489
F8	Female	20+	4/12/2014	14/3/2016	2612	466

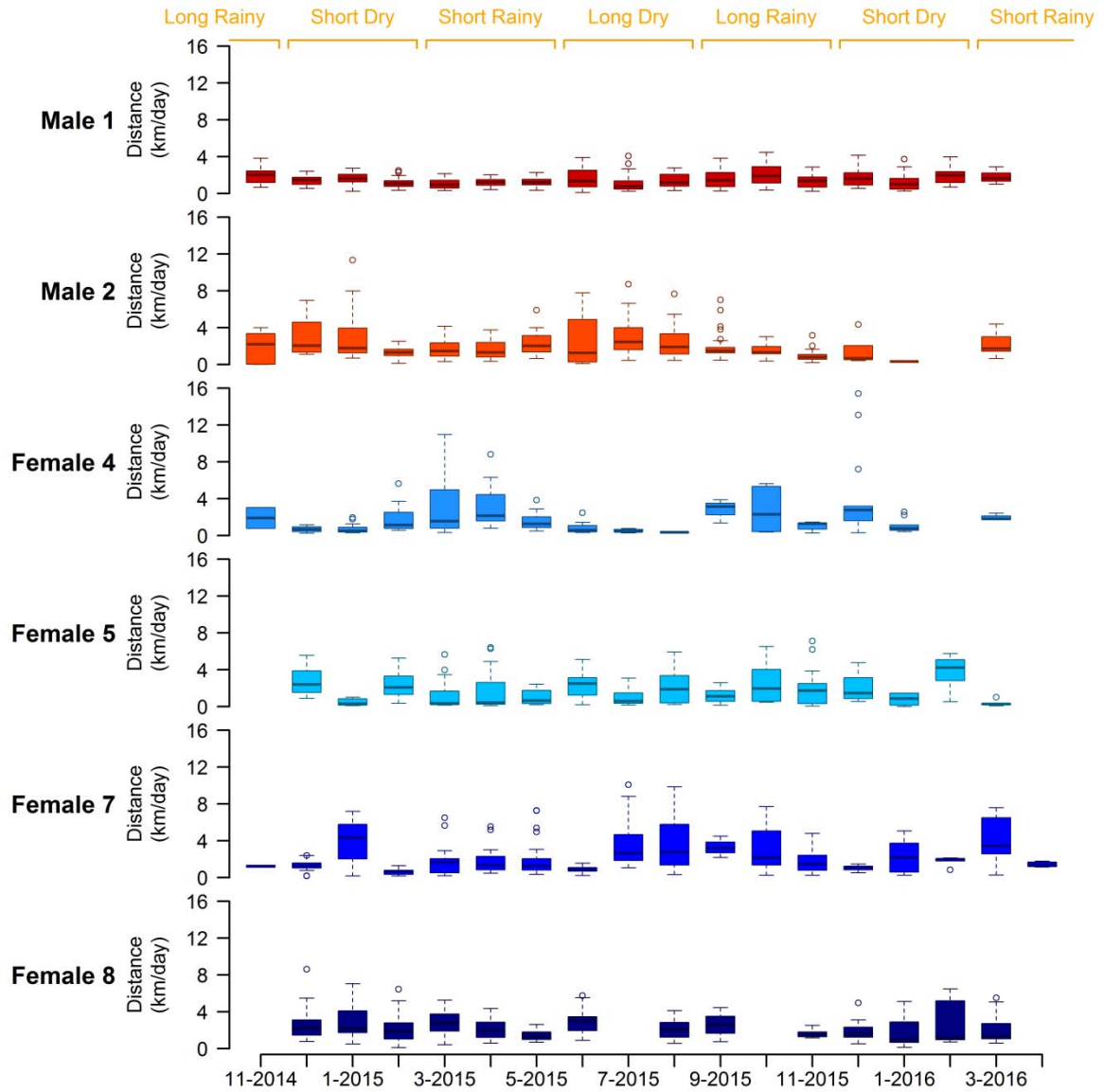
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50

51 Figure B 1. Force-directed graph based on the correlation (Pearson r^2) between the geographical,
 52 meteorological and anthropogenic variables obtained with the *R* package *qgraph* (Epskamp et al.,
 53 2012). MA corresponds to moving average, Min. to minimum and Max. to maximum.

54



55

56 Figure C 1. Box plots of the linear displacement (km) per month by each individual considered in the
 57 analysis (i.e., excluding collar #9 and collar #2). The mean tracking days per individual was 382
 58 (374–388), and the mean number of fixes per individual was 1931 (1069–5034).