- 1 Living in a landscape of fear: the impact of predation, resource availability and
- 2 habitat structure on primate range use
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36 Spatial variation in predation risk generates a "landscape of fear", with prey animals modifying their 37 distribution and behaviour in response to this variable predation risk. In systems comprised of 38 multiple predators and prey species, a key challenge is distinguishing the independent effects of 39 different predator guilds on prey responses. Here we exploit the acoustically distinct alarm calls of 40 samango monkeys (Cercopithecus mitis erythrarchus) to create a predator-specific landscape of fear 41 for eagles to assess its impact on space use within mixed regressive-spatial regressive models 42 incorporating data on resource distribution and structural characteristics of the environment. The 43 landscape of fear from eagles was the most significant determinant of samango range use, with no 44 effect of resource availability. The monkeys also selected areas of their range with higher canopies 45 and higher understory visibility, behaviour consistent with further minimising risk of predation. 46 These results contrast with those of vervet monkeys (*Chlorocebus aethiops pygerythrus*) at the same 47 site where the landscapes of fear from leopards and baboons were the most significant 48 determinants of space use. Whilst highlighting that predation risk is a key driver of primate 49 behaviour in this population, the landscapes of fear experienced by samango monkeys and vervet 50 monkeys appear to differ despite exposure to identical predator guilds. This emphasizes the 51 importance of distinguishing between the risk effects of different predators in understanding prey 52 ecology, but also that closely-related prey species may respond to these predator-specific risks in 53 different ways.

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Key words: blue monkey; Sykes monkey; *Cercopithecus albogularis; Cercopithecus aethiops*; fruit
availability; intergroup competition; geographic information system (GIS); Local Convex Hulls
(LoCoH) analysis; spatial regression.

60 Predation is a major selective force driving animal evolution (Dawkins & Krebs 1979) with almost all 61 animal species engaged in some form of predator-prey interaction (Abrams 2000). Understanding 62 how animals manage the risk of predation is thus a central issue in behavioural ecology (Quinn & 63 Cresswell 2004). Predation imposes two costs on prey individuals: the direct fitness costs of 64 mortality resulting from successful predation and the indirect costs of employing behaviours to 65 reduce mortality risks. These non-lethal effects of predators appear to impact almost every aspect 66 of prey behaviour and ecology (Caro 2005; Lima 1998; Werner & Peacor 2006). Nevertheless, these 67 risk effects are amongst the most difficult to quantify (Creel & Christianson 2008; Creel et al. 2008; 68 Hill & Dunbar 1998).

69

70 Spatial variation in risk is a key driver of non-lethal predation effects (Cresswell & Quinn 2013), 71 primarily due to the constraints this places on foraging behaviour and the subsequent impact this 72 has on competitive and trophic interactions (Creel et al. 2007; Minderman, Lind & Cresswell 2006; 73 Peckarsky & McIntosh 1998; Willems & Hill 2009). Spatial variation in perceived predation risk has 74 often been conceptualised as a 'landscape of fear' (Brown & Kotler 2004; Brown, Laundre & Gurung 75 1999; Laundré, Hernandez & Altendorf 2001), with approaches such as giving-up densities quantifying the trade-offs animals make between nutrient acquisition and the costs of predation 76 77 (Brown 1988). One of the challenges though is that these methods do not strictly measure 78 perceived predation risk (Searle, Stokes & Gordon 2008) and in multi-predator environments they do 79 not convey information on the impact of different predators on the behavioural responses of prey 80 species. This latter issue is critical, since when prey are subject to attack from several predators that 81 present different types of risk, the appropriate antipredatory responses differ between predator 82 guilds (Cresswell & Quinn 2013; Preisser, Orrock & Schmitz 2007; Shultz et al. 2004; Willems & Hill 83 2009). As a consequence, to understand how prey manage the risk of predation within their environment, the risk of predation from each predator guild must be quantified independently. 84

85 Nevertheless, there is a significant body of evidence to suggest that animals trade-off food

availability and predation risk in habitat choice (<u>Cowlishaw 1997</u>; <u>Fortin & Fortin 2009</u>; <u>Willems & Hill</u>
<u>2009</u>). The landscape of fear is thus a powerful concept in animal ecology and has been suggested
to be the key landscape within an animal's environment (<u>Brown & Kotler 2004</u>). To test this
assertion, however, methods are required that exclusively reflect perceived predation risk and
distinguish between predator-specific predation risk in determining prey behaviour.

91

92 In a novel approach, Willems and Hill (2009) showed that predator-specific landscapes of fear could 93 be constructed on the basis of vervet monkey (Chlorocebus aethiops pygerythrus) alarm calls. 94 Vervet monkeys at their South African field site were predated upon by leopard (*Panthera pardus*), 95 African crowned eagle (Stephanoaetus coronatus), chacma baboon (Papio ursinus) and African rock 96 python (Python sebae). When predator-specific landscapes of fear were combined with data on 97 resource distribution in a spatially explicit model, the ranging behaviour of the study group could be 98 interpreted as an adaptive response to the spatial availability of resources and the perceived risk of 99 predation by some, but not all, predators (Willems and Hill 2009). The landscapes of fear for baboon 100 and leopard were negatively associated with the group's utilisation distribution indicating that the 101 monkeys avoided areas of high perceived predation risk by these two predators. Furthermore, the 102 fear effects exceeded those of local resource availability in determining range use. In contrast, the 103 spatial distribution and local frequency of alarm responses to eagles and snakes did not significantly 104 affect range use. This highlighted the value of their framework in distinguishing between the effects 105 of different predators in studies of predator-prey ecology in multi-predator environments. Willems 106 and Hill (2009) also noted the potential for integrating additional variables such as the structural 107 characteristics of a habitat and the utilisation distributions of predators and neighbouring groups 108 into their modelling approach and advocated this as an avenue for future research.

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Here we apply the framework of <u>Willems and Hill (2009</u>) to a population of samango monkeys
 (*Cercopithecus mitis erythrarchus*) inhabiting the same multi-predator environment as in the original

112 vervet study. Initially we directly replicate the modelling approach and variable selection of Willems 113 and Hill (2009) to facilitate a direct comparison of our results to the previous models on vervet 114 monkeys. In doing so, we assess the impact that substrate preferences (arboreality vs. semi-115 terrestriality) have on the exposure to different predator guilds and the implications of this for 116 predator-specific landscapes of fear and range use. Samango monkeys, as an arboreal species, may 117 experience different predation threats to more terrestrial species such as vervet monkeys (Lawes 118 <u>1991</u>; <u>McGraw 2002</u>), so influencing the relative importance of different predators in driving ranging 119 behaviour.

120

121 We then extend the modelling framework of <u>Willems and Hill (2009</u>) in two ways. Firstly, we replace 122 categorical habitat types with continuous spatial measures of resource availability and structural 123 characteristics of habitats (e.g. canopy height, habitat visibility) to investigate the key drivers of 124 habitat choice within the landscape of fear. Samango monkeys have been observed to spend the 125 majority of their time high in the canopy (Thomas 1991), with the ground perceived as higher risk 126 (Emerson, Brown & Linden 2011), suggesting that areas of tall canopy will be preferred. Similarly, 127 habitats with dense understory vegetation may provide cover for terrestrial ambush predators (du 128 Bothma & Le Riche 1986) while areas of high visibility may increase an individual's ability to monitor 129 threats from predators or competitors (Cowlishaw 1994; Hill & Weingrill 2007; Jaffe & Isbell 2009). 130 For example, vervet monkeys have been shown to decrease vigilance in high visibility areas 131 (Chapman 1985; Enstam & Isbell 2002). Samango monkeys are thus predicted to prefer higher 132 visibility habitats. Since canopy height and visibility will vary independently between habitat types, 133 consideration of the spatial variation in these parameters should be more informative than broad 134 habitat classifications. Secondly, we explore the impact of competition with neighbouring groups. 135 Intraspecific competition and the active avoidance of neighbouring groups has long been recognized 136 as a significant factor shaping space-use strategies and movement decisions in many species (Gibson

- 137 <u>& Koenig 2012</u>; Markham et al. 2013), and yet how intraspecific competition, predation risk and
- 138 resource availability interact in determining spatial range use is largely unknown.
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- 140
- 141 **METHODS**
- 142
- 143 Study species and field site
- 144 Samango monkeys are medium sized (adult females ~4.4kg, adult males ~7.6kg: <u>Harvey, Martin and</u>
- 145 <u>Clutton-Brock (1987</u>)) arboreal, diurnal guenons. They form single-male, multi-female groups (<u>Henzi</u>
- 146 <u>& Lawes 1987; Rudran 1978</u>), with group sizes ranging from 4-65 (<u>Beeson et al. 1996</u>; <u>Butynski 1990</u>;
- 147 Houle, Chapman & Vickery 2010; Lawes, Cords & Lehn 2013; Smith, Link & Cords 2008). Samango
- 148 monkeys are primarily frugivorous but supplement their diets with leaves, insects and flowers
- 149 (Coleman 2013; Lawes 1991; Lawes, Henzi & Perrin 1990). Like vervet monkeys, samango monkeys
- 150 have acoustically distinct alarm calls for different predator guilds which can be differentiated by
- 151 human observers (Brown 1989; Papworth et al. 2008).
- 152
- 153 Research was conducted at the Lajuma Research Centre, located in the Soutpansberg Mountains, 154 Limpopo Province, South Africa (23°02'23"S, 29°26'05"E). Substantial local variation in abiotic 155 factors such as elevation and water availability results in a variety of microclimates which are able to 156 support a substantial diversity of both flora and fauna (Brock, Nortje & Gaigher 2003; Willems 2007). 157 The study area has natural fragments of tall forest (10-20m height) occurring amongst areas of 158 natural short forest (5-10m height). Local climate is classified as temperate/mesothermal, with cool 159 dry winters from April-September and warm to hot wet summers from October-March (Willems 160 2007). Mean annual temperature on site averages 17.1°C, with a mean annual rainfall of 724mm 161 (Willems, Barton & Hill 2009). On site, samango monkeys are sympatric with vervet monkeys, 162 chacma baboon, thick-tailed galago (Otolemur crassicaudatus) and southern lesser bushbaby

163 (Galago moholi). Potential predators include leopard, crowned eagle, African black eagle (Aquila

164 verreauxii) and African rock python. Venomous snakes, including black mamba (Dendroaspis

165 *polylepis*), puff adder (*Bitis arietans*) and Mozambique spitting cobra (*Naja mossambica*), whilst not

actively preying on samango monkeys, still pose potential mortality threats and may affect range

167 use.

168

#### 169 Behavioural data

170 A well-habituated group of approximately 40 samango monkeys was observed over a 16 month 171 period (September 2009-December 2010). We restrict the data presented here to that collected during the final 12 months (January-December 2010) in order to confine the analyses to a single 172 173 annual cycle. Behavioural data were collected over eight full follow days per month (totalling 96 174 days), with a successful day defined as following the group from dawn to dusk without losing 175 audiovisual contact for more than a total of 60 minutes. Study days ranged from approximately 11.5-176 14 hours depending upon season. Data were collected on a palmtop (Sony Clie SL-10) with 177 behavioural data collection software (Pendragon Forms 4.0; Pendragon Software, Libertyville, Illinois, USA) and a GPS (Garmin GPS 60CSX; Garmin, Olathe, Kansas, USA) continually recording 178 179 location. Supplementary data were collected in paper notebooks.

180

181 To determine perceived predation risk, the time, location and details of alarm calls were recorded on 182 an all-occurrence basis. There were a total of 131 alarm calls across the study (eagle=59, snake=3, 183 other=7, unknown=62) with no confirmed leopard-specific alarm vocalisations. Since a minimum of 184 10 observations is required for the statistical techniques used here (Borger et al. 2006), only eagle 185 alarm calls were further investigated. Eleven of the 59 eagle alarm calls were accompanied by an eagle sighting, leaving 48 potentially "false" alarm calls; these calls still express the monkeys' 186 187 perception of eagle risk, however, and are equally as informative as when a predator sighting was 188 confirmed (Willems & Hill 2009).

Intergroup encounters were defined as the study group being within visual range of another
samango group, with the time, location and details of all such encounters recorded on an alloccurrence basis. There were a total of 41 inter-group encounters of varying antagonism over the
course of the study.

194

#### 195 Environmental data

The study area was separated into eight distinct habitat types: tall forest, short forest, riverine
forest, wetland, open/closed mountain bushveld and open/closed rocky mountain bushveld (based
on criteria in <u>Mucina and Rutherford (2006)</u>) (Figure 1a). All water sources available for a minimum
of one month during the study were recorded using GPS. The locations of sleeping sites were
recorded using the final GPS location of the day from the behavioural data.

201

202 Food availability was calculated from phenological transects and random quadrat sampling; these 203 methods are the most efficient for determining all types of density-related features (Southwood & 204 Henderson 2000). A series of phenological transects were established, with 10 mature trees from 24 205 potential food species then selected and tagged to ensure an even coverage of each species 206 throughout the monkey home range as well as a range of tree sizes. All trees were monitored 207 monthly for height, crown diameter at the widest point, number of leaves, percentage mature 208 leaves, number of flowers, number of fruits and percentage ripeness (unripe/ripe/overripe). Where 209 numbers of items were too large to count, estimates were made for a single branch or section and 210 then scaled up to the size of the tree. For the purposes of this analysis, food availability estimates 211 focussed on fruit availability due to samango monkeys' mainly fruigivorous diet (Lawes 1991), with 212 eight tree species, accounting for more than 67.1% of the total fruit intake (Coleman 2013) extracted from the phenological data set. For each species, linear regression analysis was used to derive 213 214 equations that expressed annual food availability for each species as a function of tree height and/or crown diameter. These equations were then used to estimate food availability for trees of known
height and crown diameter from the quadrat sampling (below). Further details of the equations used
are given in <u>Coleman (2013)</u>.

218

219 Quadrat sampling was used to calculate food availability and habitat structure within habitat types 220 and across the home range. Each month, a minimum of 100 5m x 5m quadrats were randomly 221 selected throughout the monkey home range using the ArcGIS add-on Hawth's Tools (Beyer 2004), 222 with a total of 1268 quadrats sampled across the study. Within each quadrat, all trees with a 223 diameter greater than 10cm at 1m were identified and measured for height and crown diameter. 224 Estimated total fruits per tree were then calculated on the basis of the species-specific equations derived from the transects and converted to fruit volume based on average fruit size for each 225 226 species (based on measurements in <u>Coates-Palgrave (1996</u>); acacia pods were given a nominal 227 thickness of 1mm). Total fruit volume per 25m<sup>2</sup> guadrat was then calculated for the 1268 guadrats 228 within the home range.

229

Mean tree height per quadrat was used to determine canopy height for each of the 1268 quadrats sampled. To estimate habitat visibility, a percentage understory visibility measurement was made for 632 quadrats using a 0.8m x 0.8m checkerboard (divided into 10cm squares). An observer was located at the northwest corner of the quadrat, and a field assistant then held the checkerboard at a height of two metres and a distance of five metres in each of the four cardinal points directions. The observer recorded the proportion of the grid visible in each direction and understory visibility was calculated for each quadrat as the mean of these four measurements.

237

#### 238 Spatial landscapes

All data were imported into ArcGIS 10 (Environmental Systems Research Institute, California), with
 data projected into the Universal Transverse Mercator coordinate system (datum, WGS 1984; zone,

35 S) and the cell size of all output rasters set to 3m, consistent with GPS accuracy in the field. A
series of layers were then computed to characterise the spatial distribution of the different
parameters.

244

245 Utilisation Distribution

246 Local Convex Hulls (LoCoH) analysis (Getz et al. 2007; Getz & Wilmers 2004) was used to determine 247 utilisation distribution. This method was chosen over kernel density estimation due to its superior 248 convergence properties and ability to cope better with hard boundaries such as cliffs and rivers and 249 clumped data points (Getz et al. 2007; Hemson et al. 2005; Ryan, Knechtel & Getz 2006; Silverman 250 1986; Steury et al. 2010). Ranging data were filtered to give a location point every ten minutes, providing 6912 points for analysis. An adaptive LoCoH utilisation distribution (Getz & Wilmers 2004) 251 252 was created using R (Version 2.13: <u>R Core DevelopmentTeam (2011</u>)). The adaptive LoCoH method 253 is one of three variations of LoCoH analysis, and enables smaller convex hulls to arise in higher usage 254 areas, allowing more detailed information in areas of clumped data (Getz et al. 2007), such as 255 around sleeping sites or key food resources that are used repeatedly or for long periods of time. For 256 adaptive LoCoH analysis it is suggested that the widest point between two locations is used as the 257 value a in order to ensure the correct formation of the 100% isopleth (Getz et al. 2007); here a was 258 set to 1329 metres with the utilisation distribution calculated in 1% isopleths (Figure 2). 259

260 *Resource availability and habitat structure* 

Initially, estimates of mean food availability, canopy height and visibility were computed from the
quadrats sampled for each of the eight habitat types (Table 1) and linked to the habitat layer within
GIS. To investigate the effect of the spatial variation in environmental parameters more precisely,
landscapes of fruit availability, canopy height and understory visibility were then constructed.
Quadrat data for each of these variables were interpolated using kriging (Cressie 1990) to generate
the landscapes (Figure 1b-d). The search radii for kriging were calculated based on the number of

267 points achieving minimum root mean squared error (food resources: 45; canopy height: 45; visibility:

268 60) (Salih et al. 2002). Finally, access to water and sleeping sites were expressed as the shortest

269 Euclidean distances to the nearest water source and confirmed sleeping tree taken from the

270 behavioural data (Figure 1e-f). All layers were clipped to the utilisation distribution.

271

- 272 Predation risk and intergroup encounter risk
- A fixed kernel density estimation (<u>Silverman 1986</u>) was employed to create a density distribution of

274 eagle alarm vocalisations and inter-group encounters. A PLUGIN bandwidth parameterisation was

used since on smaller samples it has been shown to have less variability and outperform least-

276 squares cross validation (LSCV: (<u>Gitzen, Millspaugh & Kernohan 2006</u>; <u>Lichti & Swihart 2011</u>).

277 Following <u>Willems and Hill (2009</u>), measures of the probability of an alarm response or group

278 encounter occurring at each point per unit of time the monkeys spent there were then calculated by

279 dividing the kernel density estimations by the utilisation distribution within ArcGIS to create the

280 landscapes of fear for eagles and inter-group competition (Figure 1g-h).

281

#### 282 Statistical analysis

283 Following Willems and Hill (2009), a random set of 1000 points from within the home range were 284 selected using Geospatial Modelling Environment (Version 0.5.5 Beta: Beyer (2011)) and parameter 285 values were extracted. The spatial patterns within these data were assessed by inspection of 286 correlograms and Moran's / values calculated in ArcGIS 10.0 (Figure 3). Two mixed regressive-spatial 287 regressive (or lagged predictor) models were conducted to determine the combined effect of the 288 predictor variables on the intensity of space whilst also accounting for spatial autocorrelation 289 (Willems & Hill 2009). Model A incorporated habitat types (recoded into seven dummy variables) 290 and included perceived eagle predation risk, distance from water and sleeping sites as separate 291 predictor variables. Model B replaced habitat types with fruit availability, canopy height and 292 understory visibility and included perceived eagle predation risk, distance from water and sleeping

293 sites and inter-group encounter risk as separate predictor variables. Models were selected based on 294 the AIC selection method suggested by Richards (2008). Initially all models with an  $\Delta$ -value  $\leq$  6 were 295 selected with a more complex model only chosen if its AIC value was less than the AIC value of all 296 the simpler models within which it was nested. Visual inspection of the residuals from the models 297 confirmed the data were close to normally distributed (Figure A1). Spatial analyses were conducted 298 using the package Spatial Analysis in Macroecology 4.0 (Rangel, Diniz-Filho & Bini 2010). 299 300 RESULTS 301 302 303 Two mixed regressive-spatial regressive models were used in order to determine the extent to which 304 the variation in intensity of space use could be ascribed to the simultaneous effects of all 305 investigated parameter. In Model A, containing the different habitat types, the full model included a 306 non-significant effect of distance to water (Table A1) and water availability was dropped following 307 AIC selection. The final model indicated significant negative relationships between intensity of range 308 use and both perceived eagle predation and distance to sleeping sites (Table 2), with the 309 standardised regression parameters and t values identifying eagle risk as the most important 310 variable in the model. Significant relationships were also revealed with six of the seven habitat 311 variables; range use intensity increased in tall forest, with highly significant negative relationships for 312 riverine forest, open and closed mountain bushveld and open and closed rocky mountain bushveld. 313 No significant relationship existed for short forest despite this habitat containing the highest food 314 availability within the monkey home range (Table 1). The one preferred habitat, tall forest, 315 contained the second highest mean canopy height of the habitats available within the home range. 316 For Model B, the habitat types were replaced with separate landscapes of food availability, 317

318 understory visibility and canopy height, and intergroup encounter risk was also incorporated into the

analysis. The full model contained non-significant effects of intergroup encounter risk and food
availability (Table A2) and these were dropped following AIC selection. The best model contained
four parameters (Table 3). As for Model A, range use intensity was significantly negatively related to
perceived eagle predation risk and distance to sleeping sites, with the standardised regression
parameters and t values indicating that the landscape of fear of eagles had the strongest effect.
Canopy height and understory visibility were positively related to the utilisation distribution with the
monkeys preferring areas of tall trees and high visibility.

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327

#### 328 DISCUSSION

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330 The non-lethal effects of predators are increasingly recognised as one of the most significant 331 constraints on prey behaviour and yet these effects are amongst the most difficult to quantify. Here 332 we extended the spatially explicit models of Willems and Hill (2009) in order to explore the 333 significance of arboreal substrate use and fine-grained environmental variables for understanding 334 primate ranging behaviour. The key driver of samango monkey range use was spatial variation in the 335 risk of predation from eagles in both models. In our initial model based on habitat types, strong 336 relationships were found with almost every habitat type, although the relationships were not 337 consistent with habitat selection based on food availability or canopy structure. Strikingly, food 338 availability was also not a significant parameter in our model based on continuous environmental 339 landscapes, with canopy height and understory visibility identified as significant parameters. 340 Nevertheless, given that selection for tall trees and the avoidance of areas with low understory 341 visibility is consistent with a predator avoidance strategy (Emerson et al. 2011; Jaffe & Isbell 2009), 342 these results highlight the significance of the landscape of fear as a key determinant of animal space 343 use and behaviour.

345 Eagle predation risk was the strongest predictor of samango monkey space use in both models, with 346 the strong negative effects suggesting that the eagles posed enough danger that samangos avoided 347 areas considered high risk. While supporting the significance of predation as the principal parameter 348 driving range use decisions, the significance of eagle risk in our study does contrast with Willems and 349 Hill (2009) who found no significant relationship between an eagle landscape of fear and vervet 350 ranging behaviour in the same habitat. Willems and Hill (2009) suggested that the eagles' ability to 351 range over large hunting areas resulted in a relatively even distribution of predation risk across a 352 primate's home range, such that adjustments in vertical space use by prey were a more effective 353 antipredation strategy for this predator guild. Although we cannot rule out the fact that the 354 differences in our results could emerge from us not formally incorporating landscapes of fear from 355 leopards and snakes in our models, our results do suggest that predation risk from eagles does 356 indeed produce a horizontal landscape of fear.

357

358 The two highest areas of perceived eagle risk in our study were close to two known nesting sites of 359 breeding pairs of eagles; in the northwest a crowned eagle nest and in the east a black eagle pair 360 (Figure 1g). As a consequence, direct encounters in these areas may have been more frequent. 361 Many eagles prefer to hunt from a perched position high in the canopy, especially in areas of 362 relatively dense vegetation or high density of prey species (Garrett, Watson & Anthony 1993; Shultz 363 2001; Valdez & Osborn 2004). Since samangos select areas of tall continuous forest canopy this is 364 likely to increase their risk of encounter with eagles, particularly in comparison to vervet monkeys. 365 As a consequence samango monkeys may be exposed to greater risk of predation from raptors in 366 comparison to semi-terrestrial vervet monkeys, potentially accounting for the differences in 367 response to this predator guild between the two species and the existence of a strong eagle 368 landscape of fear in the arboreal species.

370 We recorded no evidence of leopard-specific alarm calls by samango moneys, despite the landscape 371 of fear from leopards being the strongest predictor of vervet monkey space use on site (Willems & 372 Hill 2009). Papworth et al. (2008) identified a leopard-specific alarm vocalisation in a population of 373 samango monkeys in Uganda, although the male 'pyow' call has also been suggested to be a 374 territory call (Cords 1987; Marler 1973) or having the characteristics of a general alarm call (Brown 375 <u>1989</u>). Further work is therefore required to determine the validity of acoustically distinct calls for 376 different predator guilds in samango monkeys. Nevertheless, dietary data from the high-density 377 leopard population at Lajuma (10.73 leopards per 100 km<sup>2</sup>: Chase Grey, Kent and Hill (in press)) 378 indicate that samangos are less frequent prey compared to vervet monkeys (vervets 12.2%; 379 samangos 2.1%: Chase-Grey (2011)). This suggests that the samango monkeys are successful in 380 avoiding predation from this species at our site. Nevertheless, further work is clearly needed to 381 determine the extent to which fear of terrestrial predators impacts on space use in samango 382 monkeys. Similarly, while the small number of snake alarm calls probably reflects the fact that most 383 on-site species of snake were not true predators, additional data are needed to explore the 384 landscape of fear from snakes further.

385

386 In combination with the results of Willems and Hill (2009), the current study suggests that predation 387 risk is a key driver of ranging behaviour for both samango monkeys and vervet monkeys. It is 388 important to note though, that the landscapes of fear experienced by the two species differ 389 markedly in this population, despite exposure to identical predator guilds. The landscape of fear 390 from eagles was the most significant influence on range use for samango monkeys, while fear of 391 leopards and baboons was the key driver for the sympatric vervet monkeys (Willems & Hill 2009). 392 Although further work is needed to assess the importance of leopards and snakes to samangos, the 393 current findings highlight the importance of distinguishing between the risk effects of different 394 predators in understanding prey ecology (Cresswell & Quinn 2013; Morosinotto, Thomson &

395 Korpimaki 2010; Preisser et al. 2007), since differences in the anti-predatory responses of species
 396 may only be apparent when risk is assessed on a predator-specific basis.

397

398 Food availability had no significant effect on samango monkey space use, either at the level of 399 habitat type or when food availability was quantified as a spatial landscape. In contrast, canopy 400 height was the most significant habitat parameter in the second model, with areas of high 401 understory visibility also significantly preferred. Such relationships were obscured in model A, 402 however, where habitat classifications integrate information on food availability and vegetation 403 structural into a single metric. This highlights the value replacing categorical habitat types with 404 continuous landscapes of environmental variables since in model A it was difficult to infer the 405 underlying habitat characteristics driving selection for particular habitat types. Although the 406 methods employed in our second analysis are more labour intensive, and so may not be appropriate 407 in all situations, future studies should adopt approaches that ensure environmental parameters are 408 recorded at spatial scales appropriate to the ranging data and utilisation distribution.

409

410 Preference for tall forest is unsurprising for an arboreal species, but model B highlights that areas of 411 relatively taller canopy were being selected within the forest habitats with the monkeys showing a 412 strong preference for tree height. The findings are consistent with previous observations that 413 samango monkeys spend the majority of their time high in the canopy (Thomas 1991), behaviour 414 that has previously been reported to decrease predation risk from terrestrial predators (Hart, 415 Katembo & Punga 1996). An experimental study based on giving up densities highlighted that the 416 ground was perceived as higher risk by samangos in this population (Emerson et al. 2011) suggesting 417 that terrestrial predators may be an important driver of habitat selection for the monkeys even if 418 this is not reflected in the distribution of alarm vocalisations. The avoidance of areas with low 419 visibility and dense understory vegetation that may conceal terrestrial ambush predators (du 420 Bothma & Le Riche 1986) is consistent with this interpretation, but further research is needed to

421 determine the importance of terrestrial predators in shaping the range use of these arboreal422 primates.

423

424 The selection of habitats on the basis of canopy height and understory visibility appears to account 425 for the lack of independent effects of food availability within our models. Although food availability 426 correlates positively with canopy height within the study area, low visibility habitats are also high in 427 fruit biomass such that interactions between these habitat effects may mask any direct influence of 428 resource availability on ranging (Coleman 2013). It is important to note though that in using annual 429 fruit availability in our spatial models we may not detect the more subtle seasonal effects resulting 430 from samangos exploiting areas of short-term high fruit availability (Willems et al. 2009). Samango 431 monkeys also possess cheek pouches that are thought to play an important role in minimising 432 exposure to predators (Smith et al. 2008). Samangos may thus minimise time in areas of high food 433 availability but high predation risk by filling their cheek pouches and then moving to areas of low 434 food availability but increased safety from predators to consume the food. Further research 435 examining cheek pouch use within the landscape of fear is required to confirm this prediction. 436 437 Finally, we found no effect of intergroup encounters on samango monkey space use; however, our 438 use of annual landscapes may disguise seasonal effects (see Markham et al. (2013)). Lawes and 439 Henzi (1995) reported that 48% of inter-group encounters in samango monkeys were food related, 440 with territory defence and mate defence potential explanations for the remaining encounters. The 441 relative importance of these factors is likely to vary significantly on an annual basis, leading to 442 differing selection pressures relating to space use. In baboons, broad temporal changes in ecological 443 resources were the major predictor of how intensively group ranges overlapped, but spacing 444 increased significantly in weeks where social groups had high proportions of fertile females 445 (Markham et al. 2013). Distinguishing between mating and non-mating seasons is thus a key future

446	direction, although simultaneous monitoring of the ranging behaviour of the neighbouring groups
447	will also help to separate the independent effects of conspecific groups on range use.
448	
449	Studies of the effects of predators on the behaviour and abundance of their prey have traditionally
450	assumed that all predators have the same selective effects (Lima 2002) but the inaccuracy of this
451	assumption is increasingly recognized (Cresswell & Quinn 2013; Preisser et al. 2007; Schmitz 2008).
452	The results presented here suggest that similar prey species inhabiting the same environment do not
453	experience the same selective effects from the different predator guilds, despite being exposed to
454	the same predator community. The challenge for future work, therefore, is to determine how
455	predator diversity and hunting mode coupled with variation in prey responses shapes the dynamics
456	of prey communities ( <u>Preisser et al. 2007</u> ).
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458	
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460	
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469	REFERENCES

470 Abrams, P. A. 2000. The evolution of predator-prey interactions: Theory and evidence. *Annual*471 *Review of Ecology and Systematics*, **31**, 79-105.

- 472 Beeson, M., Tame, S., Keeming, E. & Lea, S. E. G. 1996. Food habits of guenons (Cercopithecus spp.)
- 473 in Afro-montane forest. African Journal of Ecology, 34, 202-210.
- 474 Beyer, H. L. 2004. Hawth's Analysis Tools for ArcGIS (Version 9), available at
- 475 www.spatialecology.com/htools.
- 476 Beyer, H. L. 2011. Geospatial Modelling Environment (Version 0.55 Beta), available at
- 477 www.spatialecology.com/gme
- 478 Borger, L., Franconi, N., De Michele, G., Gantz, A., Meschi, F., Manica, A., Lovari, S. & Coulson, T.
- 479 2006. Effects of sampling regime on the mean and variance of home range size estimates. Journal of 480 Animal Ecology, **75**, 1393-1405.
- Brock, A. P., Nortje, E. & Gaigher, I. 2003. Habitat preference of a local population of bushbuck in 481
- 482 the Soutpansberg Mountains, South Africa. South African Journal of Wildlife Research, 33, 131-135.
- 483 Brown, C. H. 1989. The active space of blue monkey and grey-cheeked mangabey vocalizations. 484 Animal Behaviour, 37, 1023-1034.
- 485 Brown, J. S. 1988. Patch use as an indicator of habitat preference, predation risk and competition. 486 Behavioral Ecology and Sociobiology, 22, 37-47.
- 487 Brown, J. S. & Kotler, B. P. 2004. Hazardous duty pay and the foraging cost of predation. Ecology 488 Letters, 7, 999-1014.
- 489 Brown, J. S., Laundre, J. W. & Gurung, M. 1999. The ecology of fear: Optimal foraging, game theory, 490 and trophic interactions. Journal of Mammalogy, 80, 385-399.
- 491 Butynski, T. M. 1990. Comparative ecology of blue monkeys (Cercopithecus mitis) in high-density
- 492 and low-density subpopulations. Ecological Monographs, 60, 1-26.
- 493 Caro, T. M. 2005. Antipredator defenses in birds and mammals. Chicago: University of Chicago Press.
- 494 Chapman, C. 1985. The influence of habitat on behaviour in a group of St-Kitts Green Monkeys. 495 Journal of Zoology, 206, 311-320.
- 496 Chase-Grey, J. N. 2011. Leopard population dynamics, trophy hunting and conservation in the
- 497 Soutpansberg Mountains, South Africa, Durham University.
- 498 Chase Grey, J. N., Kent, V. T. & Hill, R. A. in press. Evidence of a high density population of harvested 499 leopards in a montane environment Plos One.
- 500 Coates-Palgrave, K. 1996. Trees of Southern Africa (Second Revised Edition). Cape Town: Struik
- 501 Publishers (Pty) Ltd.
- 502 Coleman, B. T. 2013. Spatial and temporal determinants of samango monkey (Cercopithecus mitis
- 503 erythrarchus) resource acquisition and predation avoidance behaviour. PhD Thesis, Durham 504 University.
- 505 Cords, M. 1987. Mixed species association of *Cercopithecus* monkeys in the Kakamega Forest, Kenya.
- 506 University of California Publications in Zoology, **117**, 1-109.
- 507 Cowlishaw, G. 1994. Vulnerability to Predation in Baboon Populations. *Behaviour*, **131**, 293-304.
- 508 Cowlishaw, G. 1997. Trade-offs between foraging and predation risk determine habitat use in a 509
- desert baboon population. Animal Behaviour, 53, 667-686.
- 510 Creel, S. & Christianson, D. 2008. Relationships between direct predation and risk effects. Trends in 511 *Ecology & Evolution*, **23**, 194-201.
- 512 Creel, S., Christianson, D., Liley, S. & Winnie, J. A., Jr. 2007. Predation risk affects reproductive
- 513 physiology and demography of elk. Science, 315, 960-960.
- 514 Creel, S., Winnie, J. A., Jr., Christianson, D. & Liley, S. 2008. Time and space in general models of
- 515 antipredator response: tests with wolves and elk. Animal Behaviour, 76, 1139-1146.
- 516 Cressie, N. 1990. The origins of kriging. *Mathematical Geology*, 22, 239-252.
- 517 Cresswell, W. & Quinn, J. L. 2013. Contrasting risks from different predators change the overall
- 518 nonlethal effects of predation risk. Behavioral Ecology, 24, 871-876.
- 519 Dawkins, R. & Krebs, J. R. 1979. Arms races between and within species. Proceedings of the Royal
- 520 Society Series B-Biological Sciences, **205**, 489-511.

- 521 du Bothma, J. & Le Riche, E. A. N. 1986. Prey preference and hunting efficiency of the Kalahari
- 522 leopard. In: Cats of the world: biology, conservation and management. (Ed. by S. D. Miller & D. D. 523 Everett), pp. 389-414. Washington DC: National Wildlife Federation.
- 524 Emerson, S. E., Brown, J. S. & Linden, J. D. 2011. Identifying Sykes' monkeys', Cercopithecus
- 525 albogularis erythrarchus, axes of fear through patch use. Animal Behaviour, 81, 455-462.
- 526 Enstam, K. & Isbell, L. 2002. Changes in visibility affect ranging behavior and vigilance in vervet
- 527 monkeys (Cercopithecus aethiops). American Journal of Physical Anthropology, 67-67.
- 528 Fife-Schaw, C., Hammond, S. & Breakwell, G. M. 2006. Research methods in psychology (3rd 529 edition). London: SAGE.
- 530 Fortin, D. & Fortin, M.-E. 2009. Group-size-dependent association between food profitability,
- 531 predation risk and distribution of free-ranging bison. Animal Behaviour, 78, 887-892.
- 532 Garrett, M. G., Watson, J. W. & Anthony, R. G. 1993. Bald eagle home range and habitat use in the 533 Columbia River estuary. Journal of Wildlife Management, 57, 19-27.
- 534 Getz, W. M., Fortmann-Roe, S., Cross, P. C., Lyons, A. J., Ryan, S. J. & Wilmers, C. C. 2007. LoCoH:
- 535 Nonparameteric kernel methods for constructing home ranges and utilization distributions. Plos 536 One, **2**.
- 537 Getz, W. M. & Wilmers, C. C. 2004. A local nearest-neighbor convex-hull construction of home 538 ranges and utilization distributions. Ecography, 27, 489-505.
- 539 Gibson, L. & Koenig, A. 2012. Neighboring groups and habitat edges modulate range use in Phayre's
- 540 leaf monkeys (Trachypithecus phayrei crepusculus). Behavioral Ecology and Sociobiology, 66, 633-541 643.
- 542 Gitzen, R. A., Millspaugh, J. J. & Kernohan, B. J. 2006. Bandwidth selection for fixed-kernel analysis
- 543 of animal utilization distributions. Journal of Wildlife Management, 70, 1334-1344.
- 544 Hart, J. A., Katembo, M. & Punga, K. 1996. Diet, prey selection and ecological relations of leopard 545 and golden cat in the Ituri Forest, Zaire. African Journal of Ecology, 34, 364-379.
- 546 Harvey, P. H., Martin, R. D. & Clutton-Brock, T. H. 1987. Life histories in comparative perspective.
- 547 In: Primate societies (Ed. by B. B. Smuts), pp. 181-196. Chicago: Univ. of Chicago P.
- 548 Hemson, G., Johnson, P., South, A., Kenward, R., Ripley, R. & Macdonald, D. 2005. Are kernels the 549 mustard? Data from global positioning system (GPS) collars suggests problems for kernel home-
- 550
- range analyses with least-squares cross-validation. Journal of Animal Ecology, 74, 455-463.
- 551 Henzi, S. P. & Lawes, M. 1987. Breeding-season influxes and the behavior of adult male samango
- 552 monkeys (Cercopithecus mitis albogularis). Folia Primatologica, 48, 125-136.
- 553 Hill, R. A. & Dunbar, R. I. M. 1998. An evaluation of the roles of predation rate and predation risk as 554 selective pressures on primate grouping behaviour. Behaviour, 135, 411-430.
- 555 Hill, R. A. & Weingrill, T. 2007. Predation risk and habitat use in chacma baboons (Papio hamadryas 556 ursinus). In: Primate anti-predator strategies (Ed. by S. Gursky & K. A. I. Nekaris), pp. 339-354. New 557 York: Springer.
- Houle, A., Chapman, C. A. & Vickery, W. L. 2010. Intratree vertical variation of fruit density and the 558 559 nature of contest competition in frugivores. Behavioral Ecology and Sociobiology, 64, 429-441.
- 560 Jaffe, K. E. & Isbell, L. A. 2009. After the fire: Benefits of reduced ground cover for vervet monkeys 561 (Cercopithecus aethiops). American Journal of Primatology, 71, 252-260.
- 562 Laundré, J. W., Hernandez, L. & Altendorf, K. B. 2001. Wolves, elk, and bison: reestablishing the
- "landscape of fear" in Yellowstone National Park, USA. Canadian Journal of Zoology-Revue 563
- 564 *Canadienne De Zoologie*, **79**, 1401-1409.
- 565 Lawes, M. J. 1991. Diet of samango monkeys (Cercopithecus mitis erythrarchus) in the Cape Vidal 566 Dune Forest, South-Africa. Journal of Zoology, 224, 149-173.
- Lawes, M. J., Cords, M. & Lehn, C. 2013. Cercopithecus mitis profile. In: Primates of Africa volume II, 567
- 568 The mammals of Africa (Ed. by J. Kingdon & J. Kalina), pp. 354-362. Berkeley: University of California
- 569 Press.
- 570 Lawes, M. J. & Henzi, S. P. 1995. Intergroup encounters in blue monkeys: How territorial must a
- 571 territorial species be. Animal Behaviour, 49, 240-243.

- 572 Lawes, M. J., Henzi, S. P. & Perrin, M. R. 1990. Diet and feeding-behavior of samango monkeys
- 573 (*Cercopithecus mitis labiatus*) in Ngoye Forest, South-Africa. *Folia Primatologica*, **54**, 57-69.
- 574 **Lichti, N. I. & Swihart, R. K.** 2011. Estimating utilization distributions with kernel versus local convex 575 hull methods. *Journal of Wildlife Management*, **75**, 413-422.
- 576 Lima, S. L. 1998. Nonlethal effects in the ecology of predator-prey interactions What are the
- 577 ecological effects of anti-predator decision-making? *BioScience*, **48**, 25-34.
- 578 Lima, S. L. 2002. Putting predators back into behavioral predator-prey interactions. *Trends in Ecology*
- 579 & Evolution, **17**, 70-75.
- 580 Markham, A. C., Guttal, V., Alberts, S. C. & Altmann, J. 2013. When good neighbors don't need
- fences: temporal landscape partitioning among baboon social groups. *Behavioral Ecology and Sociobiology*, **67**, 875-884.
- 583 Marler, P. 1973. A comparison of vocalizations of red-tailed monkeys and blue monkeys,
- 584 *Cercopithecus ascanius* and *C. mitis*, in Uganda. *Zeitschrift Fur Tierpsychologie-Journal of* 585 *Comparative Ethology*, **33**, 223-247.
- 586 **McGraw, W. S.** 2002. Diversity of guenon positional behavior. In: *The guenons: Diversity and*
- *adaptation in African monkeys* (Ed. by M. A. Glenn & M. Cords). Kluwer Academic, New York, pp113132.
- 589 Minderman, J., Lind, J. & Cresswell, W. 2006. Behaviourally mediated indirect effects: interference
   590 competition increases predation mortality in foraging redshanks. *Journal of Animal Ecology*, 75, 713 591 723.
- 592 Morosinotto, C., Thomson, R. L. & Korpimaki, E. 2010. Habitat selection as an antipredator
- behaviour in a multi-predator landscape: all enemies are not equal. *Journal of Animal Ecology*, **79**,
  327-333.
- Mucina, L. & Rutherford, M. C. 2006. The Vegetation of South Africa, Lesotho and Swaziland. South
   African National Biodiversity Institute, Pretoria.
- 597 Papworth, S., Bose, A. S., Barker, J., Schel, A. M. & Zuberbuhler, K. 2008. Male blue monkeys alarm
- call in response to danger experienced by others. *Biology Letters*, 4, 472-475.
   Peckarsky, B. L. & McIntosh, A. R. 1998. Fitness and community consequences of avoiding multiple
- 600 predators. *Oecologia*, **113**, 565-576.
- 601 **Preisser, E. L., Orrock, J. L. & Schmitz, O. J.** 2007. Predator hunting mode and habitat domain alter 602 nonconsumptive effects in predator-prey interactions. *Ecology*, **88**, 2744-2751.
- 603 **Quinn, J. L. & Cresswell, W.** 2004. Predator hunting behaviour and prey vulnerability. *Journal of* 604 *Animal Ecology*, **73**, 143-154.
- Rangel, T. F. L. V. B., Diniz-Filho, J. A. F. & Bini, L. M. 2010. SAM: a comprehensive application for
   Spatial Analysis in Macroecology. *Ecography*, 33.
- 607 **Richards, S. A.** 2008. Dealing with overdispersed count data in applied ecology. *Journal of Applied* 608 *Ecology*, **45**, 218-227.
- Rudran, R. 1978. Socioecology of the blue monkeys of the Kibale Forest, Uganda. Smithsonian
   Contributions to Zoology, 249, 1-88.
- 611 **Ryan, S. J., Knechtel, C. U. & Getz, W. M.** 2006. Range and habitat selection of African buffalo in 612 South Africa. *Journal of Wildlife Management*, **70**, 764-776.
- 613 Salih, I. M., Petterson, H. B. L., Sivertun, A. & Lund, E. 2002. Spatial correlation between radon in
- 614 groundwater and bedrock uranium: GIS and geostistical analyses. *Journal of Spatial Hydrology*, **2**, 1-615 10.
- 616 Schmitz, O. J. 2008. Effects of predator hunting mode on grassland ecosystem function. *Science*, **319**,
  617 952-954.
- 618 Searle, K. R., Stokes, C. J. & Gordon, I. J. 2008. When foraging and fear meet: using foraging
- hierarchies to inform assessments of landscapes of fear. *Behavioral Ecology*, **19**, 475-482.
- 620 Shultz, S. 2001. Notes on interactions between monkeys and African crowned eagles in Tai National
- 621 Park, Ivory Coast. *Folia Primatologica*, **72**, 248-250.

- 622 Shultz, S., Noe, R., McGraw, W. S. & Dunbar, R. I. M. 2004. A community-level evaluation of the
- impact of prey behavioural and ecological characteristics on predator diet composition. *Proceedings* of the Royal Society of London Series B-Biological Sciences, **271**, 725-732.
- 625 **Silverman, B. W.** 1986. *Density Estimation for Statistics and Data Analysis*. London: Chapman and 626 Hall.
- 627 Smith, L. W., Link, A. & Cords, M. 2008. Cheek pouch use, predation risk, and feeding competition in
- blue monkeys (*Cercopithecus mitis stuhlmanni*). *American Journal of Physical Anthropology*, **137**,
  334-341.
- 630 Southwood, T. R. E. & Henderson, P. A. 2000. *Ecological Methds*, 3rd edn. Oxford: Blackwell Science.
- 631 Steury, T. D., McCarthy, J. E., Roth, T. C., Lima, S. L. & Murray, D. L. 2010. Evaluation of Root-n
- Bandwidth Selectors for Kernel Density Estimation. *Journal of Wildlife Management*, **74**, 539-548.
- Team, R. D. C. 2011. R: A language and environment for statistical computing. Vienna, Austria.: R
   Foundation for Statistical Computing.
- 635 **Thomas, S. C.** 1991. Population densities and patterns of habitat use among anthropoid primates of 636 the Ituri Forest, Zaire. *Biotropica*, **23**, 68-83.
- 637 Valdez, U. & Osborn, S. 2004. Observations on the ecology of the black-and-chestnut eagle
- 638 (Oroaetus isidori) in a montane forest of southeastern Peru. Ornitologia Neotropical, 15, 31-40.
- 639 Werner, E. E. & Peacor, S. D. 2006. Lethal and nonlethal predator effects on an herbivore guild
- 640 mediated by system productivity. *Ecology*, **87**, 347-361.
- 641 Willems, E. P. 2007. From space to species. PhD Thesis, University of Durham.
- 642 Willems, E. P., Barton, R. A. & Hill, R. A. 2009. Remotely sensed productivity, regional home range
- selection, and local range use by an omnivorous primate. *Behavioral Ecology*, **20**, 985-992.
- 644 Willems, E. P. & Hill, R. A. 2009. Predator-specific landscapes of fear and resource distribution:
- 645 Effects on spatial range use. *Ecology*, **90**, 546-555.
- 646

648 LIST OF FIGURES

650	Figure 1: Spatial distribution of (a) habitat types with colours denoting food availability (CM - Closed
651	Mountain; CR - Closed Rocky; OM - Open Mountain; OR - Open Rocky; RF - Riverine Forest SF - Short
652	Forest; TF - Tall Forest; W - Wetland), (b) fruit availability, (c) canopy height, (d) understory visibility,
653	(e) Euclidean distance to sleeping sites, (f) Euclidean distance to water sources, (g) landscape of fear
654	from eagles, and (h) landscape of inter-group encounter risk in the samango monkey home range.
655	
656	Figure 2: Home range and utilisation distribution of the study group (54.7ha) constructed using
657	Adaptive LoCoH analysis. The black outline indicates the edges of the home range with dark blue
658	indicating areas of high utilisation.
659	
660	Figure 3: Correlograms and global Moran's / values for all parameters. Moran's / figures around $\pm 1$
661	signify strong positive/negative autocorrelation; a figure close to 0 means no autocorrelation.
662	
663	Figure A1: Residual distributions from the spatial regressive-mixed regressive models. Model A:
664	skewness = 0.018; kurtosis = 0.274. Model B skewness = -0.103; kurtosis = -0.211. Any score ±1 is
665	considered strongly non-normally distributed (Fife-Schaw, Hammond & Breakwell 2006).
666	

**TABLE 1** Mean fruit volume per 25m<sup>2</sup>, canopy height and understory visibility in the eight habitat

668 types in the samango monkey home range.

Habitat	Mean Fruit Volume (cm <sup>3</sup> )	Mean Canopy Height (m)	Mean Visibility (%)
Short Forest	14 977	5.01	56.41
Open Mountain	13 358	3.72	63.22
Closed Rocky	12 677	3.77	58.68
Wetland	12 675	3.91	56.31
Tall Forest	10 350	5.58	58.45
Closed Mountain	10 072	4.96	66.70
<b>Riverine Forest</b>	6.261	5.82	59.65
Open Rocky	3.680	3.95	49.94

**TABLE 2** Parameter estimates and key statistics of Model A, expressing utilisation distribution as a

672	function of perceived	l eagle predation risk.	habitat type and	distance to sleeping sites.
• • =				

## 

Predictor	в	В	SE B	SE γ	t	Р
Landscape of fear						
Eagle	-0.329	-0.338	0.028	0.192	-11.583	<0.001
Habitat types						
Short Forest	0.018	0.018	0.038	0.171	0.476	0.635
Open Mountain	-0.092	-0.121	0.022	0.518	-4.158	<0.001
Closed Rocky	-0.233	-0.289	0.024	0.408	-9.796	<0.001
Tall Forest	0.091	0.083	0.046	0.118	1.998	0.046
Closed Mountain	-0.095	-0.138	0.019	0.758	-4.989	<0.001
Riverine	-0.162	-0.211	0.022	0.495	-7.369	<0.001
Open Rocky	-0.132	-0.190	0.020	0.736	-6.538	<0.001
Environmental factors						
Sleeping sites	-0.353	-0.154	0.063	0.006	-5.582	<0.001

Habitat types are presented in descending order of fruit availability. Model statistics: *N*=1000;

 $R^2$ =0.312; ρ=0.992±0.172; γ = 0.992; AIC<sub>c</sub>=8046.844. Terms: ρ, spatial autoregressive parameter; γ,

677 spatial cross-regressive parameter; *θ*, unstandardised regression parameter; B, standardised

678 regression parameter.

- 680 **TABLE 3** Parameter estimates and key statistics of Model B, expressing utilisation distribution as a
- 681 function of predator perceived eagle predation risk, understory visibility, height of trees and sleeping
- site location.

Predictor	в	В	SE B	SE γ	t	р
Landscapes of fear						
Eagle	-0.271	278	0.029	0.192	-9.322	<0.001
Environmental factors						
Understory visibility	0.173	0.109	0.048	0.027	3.561	<0.001
Height of trees	0.398	0.277	0.045	0.041	8.838	<0.001
Sleeping sites	-0.334	-0.146	0.071	0.006	-4.720	< 0.001

### 684

# 685 Model statistics: N=1000; $R^2$ =0.169; $\rho$ =0.992±0.172; $\gamma$ = 0.992; AIC<sub>c</sub>=8215.277. Terms as for Table 3.

- 687 **Table A1:** Parameter estimates and key statistics of full Model A, a mixed regressive-spatial
- 688 regressive model expressing utilisation distribution as a function of perceived eagle predation risk,
- habitat type and distance to sleeping sites and water.
- 690

Predictor	В	β	SE β	SE γ	t	Р
Landscape of fear						
Eagle	-0.329	-0.338	0.028	0.192	-11.568	<0.001
Habitat types						
Short Forest	0.018	0.018	0.039	0.171	0.478	0.633
Open Mountain	-0.092	-0.121	0.022	0.518	-4.129	<0.001
Closed Rocky	-0.233	-0.289	0.024	0.408	-9.780	<0.001
Tall Forest	0.091	0.083	0.046	0.118	1.998	0.046
Closed Mountain	-0.095	-0.138	0.019	0.758	-4.978	<0.001
Riverine	-0.162	-0.211	0.022	0.495	-7.357	<0.001
Open Rocky	-0.132	-0.190	0.020	0.736	-6.499	<0.001
Environmental factors						
Sleeping sites	-0.354	-0.155	0.066	0.006	-5.399	<0.001
Water	0.003	0.001	0.066	0.006	0.05	0.960

Habitat types are presented in descending order of fruit availability. Model statistics: *N*=1000;

693 R<sup>2</sup>=0.312; ρ=0.992±0.172; γ = 0.992; AIC<sub>c</sub>=8051.019. Terms: ρ, spatial autoregressive parameter, γ,

694 spatial cross-regressive parameter; *B*, unstandardised regression parameter; β, standardised

695 regression parameter.

- 697 **Table A2:** Parameter estimates and key statistics of the full Model B, a mixed regressive-spatial
- 698 regressive model expressing utilisation distribution as a function of perceived eagle predation risk,

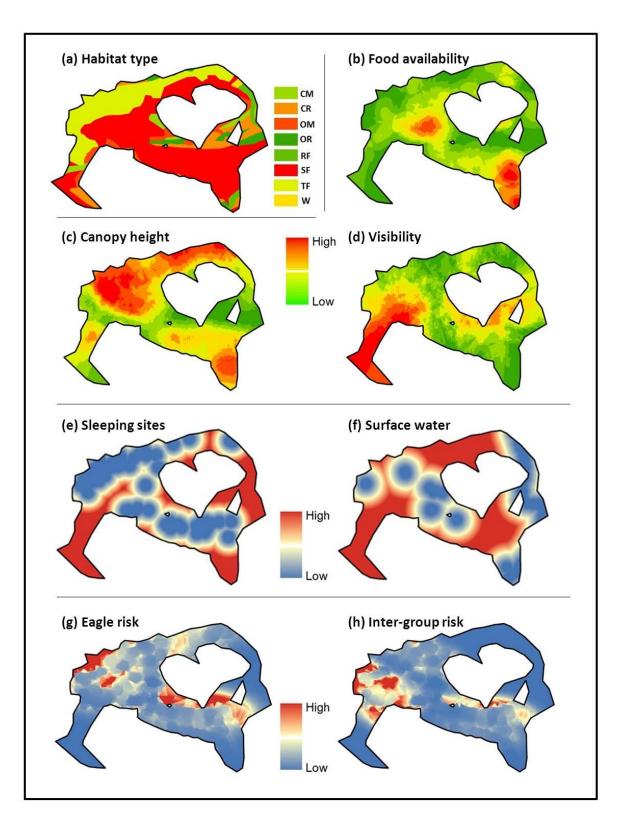
### 699 inter-group encounter risk and other environmental factors.

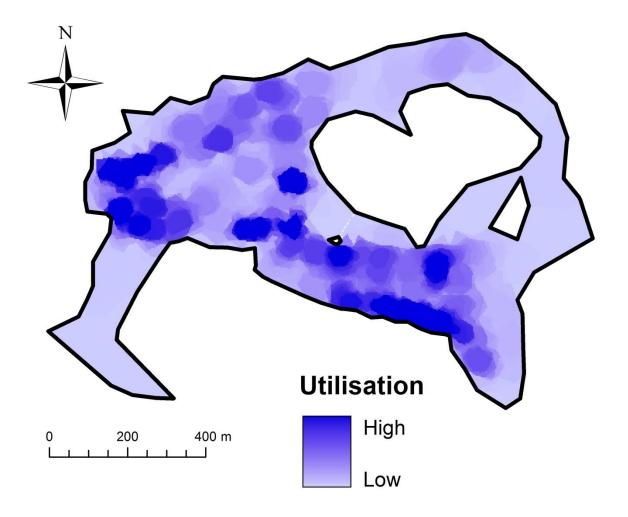
### 700

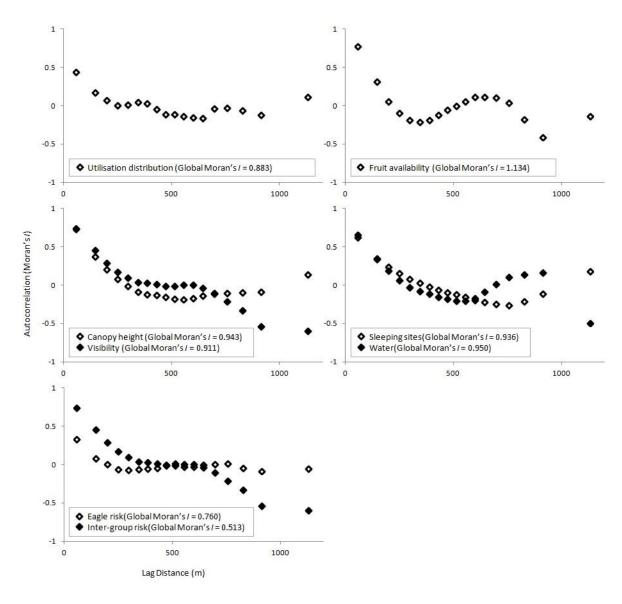
Predictor	В	β	SE β	SE γ	t	Р
Landscapes of fear						
Eagle	-0.272	-0.279	0.033	0.192	-8.309	<0.001
Inter-group	-0.018	-0.023	0.028	0.447	-0.656	0.512
Environmental factors						
Fruit availability	-0.086	-0.054	0.050	0.027	-1.711	0.087
Understory visibility	0.200	0.126	0.052	0.027	3.813	<0.001
Height of trees	0.434	0.302	0.047	0.041	9.304	<0.001
Sleeping sites	-0.266	-0.116	0.074	0.006	-3.606	<0.001
Water	-0.208	-0.089	0.077	0.006	-2.698	0.007

701

702 Model statistics: *N*=1000; R<sup>2</sup>=0.176; ρ=0.992±0.172; γ = 0.992; AIC<sub>c</sub>=8218.341. Terms as for Table A1.









# **FIGURE A1**

