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Truly sedentary? The multi-range tactic as a response to resource heterogeneity and unpredictability in a large herbivore

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Abstract:	<p>Much research on large herbivore movement has focused on the annual scale to distinguish between resident and migratory tactics, and assumes that individuals are sedentary at the within-season scale. However, apparently sedentary animals may occupy a number of sub-seasonal functional home ranges (sfHR), particularly when the environment is spatially heterogeneous and/or temporally unpredictable. The roe deer (<i>Capreolus capreolus</i>) experiences sharply contrasting environmental conditions due to its widespread distribution, but appears markedly sedentary over much of its range. Using GPS monitoring from 15 populations across Europe, we evaluated the propensity of this large herbivore to be truly sedentary at the seasonal scale in relation to variation in environmental conditions. We studied movement using net square displacement to identify the possible use of sfHR. We expected that roe deer should be less sedentary within seasons in heterogeneous and unpredictable environments, while migratory individuals should be seasonally more sedentary than residents. Our analyses revealed that, across the 15 populations, all individuals adopted a multi-range tactic, occupying between two and nine sfHR during a given season. In addition, we showed that (i) the number of sfHR was not impacted by variation in resource distribution, but decreased with increasing sfHR size; and (ii) the distance between</p>

sfHR increased with increasing heterogeneity and predictability in resource distribution, as well as with increasing sfHR size. We suggest that the multi-range tactic is likely widespread among large herbivores, allowing animals to track spatio-temporal variation in resource distribution and, thereby, to cope with changes in their local environment.

1 **Truly sedentary? The multi-range tactic as a response to resource**
2 **heterogeneity and unpredictability in a large herbivore**

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56 **Abstract**

57 Much research on large herbivore movement has focused on the annual scale to distinguish
58 between resident and migratory tactics, and assumes that individuals are sedentary at the
59 within-season scale. However, apparently sedentary animals may occupy a number of sub-
60 seasonal functional home ranges (sfHR), particularly when the environment is spatially
61 heterogeneous and/or temporally unpredictable. The roe deer (*Capreolus capreolus*)
62 experiences sharply contrasting environmental conditions due to its widespread distribution,
63 but appears markedly sedentary over much of its range. Using GPS monitoring from 15
64 populations across Europe, we evaluated the propensity of this large herbivore to be truly
65 sedentary at the seasonal scale in relation to variation in environmental conditions. We
66 studied movement using net square displacement to identify the possible use of sfHR. We
67 expected that roe deer should be less sedentary within seasons in heterogeneous and
68 unpredictable environments, while migratory individuals should be seasonally more sedentary
69 than residents. Our analyses revealed that, across the 15 populations, all individuals adopted a
70 multi-range tactic, occupying between two and nine sfHR during a given season. In addition,
71 we showed that (i) the number of sfHR was not impacted by variation in resource distribution,
72 but decreased with increasing sfHR size; and (ii) the distance between sfHR increased with
73 increasing heterogeneity and predictability in resource distribution, as well as with increasing
74 sfHR size. We suggest that the multi-range tactic is likely widespread among large herbivores,
75 allowing animals to track spatio-temporal variation in resource distribution and, thereby, to
76 cope with changes in their local environment.

77

78 **Key words:** migration, residency, sub-seasonal functional home range, spatial heterogeneity,
79 temporal predictability

80

81 **Introduction**

82 Movement is a fundamental characteristic of life which influences the survival and
83 reproduction of organisms and, more generally, individual fitness and population dynamics
84 (Turchin 1991; Revilla and Wiegand 2008). Following the Marginal Value Theorem
85 (Charnov 1976), while foraging, an individual moves within a patch (sensu Wiens 1976),
86 searching intensively for food, before leaving to search more widely for a new patch when the
87 energetic benefit of the first patch has decreased below the average value of the alternative
88 patches (Charnov 1976; Van Moorter et al. 2009). Individuals are assumed to memorise the
89 value and location of a given patch (Riotte-Lambert et al. 2015), returning to previously
90 visited patches periodically. This process leads to the emergence of a stable home range to
91 which individuals restrict their movements in order to maximise resource acquisition (Brown
92 and Orians 1970; Riotte-Lambert et al. 2015). Therefore, individuals of many species appear
93 to be sedentary at this spatio-temporal scale, occupying a stable home range over a long time
94 span (season, year). Site fidelity is widespread in the animal kingdom and has fundamental
95 consequences for ecological processes (Börger et al. 2008).

96 An individual's lifetime track is an aggregation of successive elementary units with
97 potentially different functionality (Baguette et al. 2014). Indeed, as stated by Van Moorter et
98 al. (2016) "animals do not move for the sake of changing their geographic location, but rather
99 for changing environmental conditions associated with changes in location". In general, an
100 individual decides to move in order to satisfy its requirements in terms of refuge and
101 resources (Nathan et al. 2008) which encompass changes in environmental space (Van
102 Moorter et al. 2013). For example, since favourable sites for feeding or taking refuge do not
103 necessarily occur at the same location, and since conditions vary in space and time, animals
104 have to move to cope with spatio-temporal heterogeneity in their environment (Pyke 1984;
105 Mueller and Fagan 2008; Chapman et al. 2014). As a result of variation in conditions over

106 time, an animal must shift from one suitable patch to another in order to fulfil its
107 requirements, in which case it cannot be considered truly sedentary, even at the seasonal scale
108 (Chapin et al. 1980; Barraquand and Benhamou 2008). Indeed, animals may use several
109 spatio-temporally distinct suitable units, particularly in spatially heterogeneous and/or
110 temporally unpredictable environments. While there has been considerable focus in recent
111 years on movements associated with seasonal migration, between seasonally distinct home
112 ranges (Cagnacci et al. 2011; Cagnacci et al. 2015; Peters et al. 2017), there has been
113 relatively little work on finer scale movements at the within season scale. The use of sub-
114 seasonal functional home ranges (sfHR) has previously been described in two African
115 herbivores, the sable antelope (*Hippotragus niger*) and the African savanna buffalo (*Syncerus*
116 *caffer brachyceros*) (Owen-Smith et al. 2010; Cornélis et al. 2011; Benhamou 2014).
117 However, no study has yet attempted to link the propensity of individuals to adopt this multi-
118 range tactic with spatial and temporal variation in the prevailing environmental conditions.

119 Here, we used the EURODEER database (<http://www.eurodeer.org>, see methods) to
120 analyse space use of roe deer (*Capreolus capreolus*) across widely contrasting environments,
121 from the southern part of their geographic range, in Italy, to the northern part of their range, in
122 Scandinavia. We focused on the roe deer as it is Europe's most widespread large wild
123 herbivore and is considered highly sedentary over the majority of its range (Hewison et al.
124 1998). However, this species also exhibits a considerable degree of behavioural plasticity
125 (Jepsen and Topping 2004), and is described as partially migratory in more extreme
126 environments (Cagnacci et al. 2011). Hence, we first analysed whether roe deer are truly
127 sedentary within a given season, or whether they adopt a movement tactic based on the use of
128 a series of sfHR. Secondly, we hypothesised that the propensity of an animal to adopt this
129 multi-range tactic should depend on spatio-temporal variations in environmental conditions.
130 More specifically, we predicted that individuals should be less sedentary in heterogeneous

131 compared to homogeneous environments (Mueller and Fagan 2008; Mueller et al. 2011).
132 Analogously to the nomadic movement tactic (Mueller and Fagan 2008), we also expected
133 individuals to be less sedentary in unpredictable environments, or at least in environments
134 where resources vary more markedly over time at the within-seasonal scale, compared to
135 more predictable environments. Finally, we expected that, within a given seasonal range,
136 migratory animals would be more sedentary than residents since they migrate during spring
137 and fall so that they are able to adjust their habitat use to seasonal variations in food resource
138 abundance and/or quality at this scale (Fryxell and Sinclair 1988).

139 **Material and methods**

140 *Study areas and GPS data*

141 This study was based on the database assembled by the EURODEER consortium, a data
142 sharing project to investigate the movement ecology of European deer along environmental
143 gradients (<http://eurodeer.org>, accessed on April 2016). We analysed data on 251 adult roe
144 deer (286 individual-years) from 15 study sites (see Table 1) encompassing widely
145 contrasting environmental conditions (latitude varied from 38.2°N to 60.7°N; longitude varied
146 from 0.9 °E to 23.5 °E; Fig. 1). Roe deer were captured from 2003 to 2014 using drive nets,
147 net traps or box traps depending on study site. All capture and marking procedures were done
148 in accordance with local and European animal welfare laws. Deer were equipped with GPS
149 collars programmed to obtain a GPS fix with intervals ranging from 10 minutes to 12 hours.
150 In order to standardise the data for inter-population comparisons, for each individual, we
151 restricted monitoring to the period from the 15th of February to the 15th of November, and
152 retained the two locations per day that were closest to noon and midnight.

153 *Discrimination of individual movement tactics*

154 First, we used the method proposed by Börger and Fryxell (2012), based on the net squared
155 displacement (NSD), i.e. the Euclidian distance between the starting location and all
156 subsequent locations of an individual over time (Turchin 1998), to determine each
157 individual's annual movement tactic: migration, residency or dispersal. We considered two
158 models of range residency, one with a constant NSD (the mean), and one with a linear
159 increase of NSD before reaching an asymptote; we considered one model of migration
160 including approximate dates of departure and return between seasonal ranges, and a model of
161 dispersal with an approximate date of departure (see Bunnefeld et al. 2011 and Börger and
162 Fryxell 2012 for more details on these models). In order to identify which of these models
163 best described the movement behaviour of a given individual, we used the system of non-

164 linear mixed models proposed by Börger and Fryxell (2012) which links theoretical
165 expectations to movement data. For model selection, as recommended by Börger and Fryxell
166 (2012), we retained the model with the largest concordance correlation (CC), expressing the
167 goodness of fit for each model (Huang et al. 2009). Because the assigned movement tactic
168 using this method did not always closely fit the data, we also visually examined the NSD
169 trajectories to determine each individual's annual movement tactic by eye (Bischof et al.
170 2012). We based our visual classification on the patterns of NSD typically observed for
171 migratory individuals, residents and dispersers, following Börger and Fryxell (2012). That is,
172 we assumed that when the NSD was relatively constant or increased linearly initially before
173 rapidly reaching an asymptote, the individual was resident. When the NSD was constant
174 before increasing rapidly during spring to reach a plateau during summer, then decreased
175 during fall, returning to its initial value, we assumed that the individual was migratory.
176 Finally, when the NSD was constant before increasing rapidly to reach a plateau with no
177 further increase or decrease, we assumed that the individual had dispersed. We then verified
178 that individuals which were classified as dispersers did not return to their point of departure
179 during subsequent monitoring, after the 15th of November. If they did (53 of 65 animals
180 originally classified as dispersers), these individuals were considered as migratory. We
181 excluded the remaining dispersers (N = 12) from subsequent analyses as movement patterns
182 during dispersal are governed by different ultimate causes than those involved in range
183 residency or migration (Bowler and Benton 2005; Chapman et al. 2014). After visual
184 reclassification, our data set included 193 residents and 93 migratory individuals. Note that
185 subsequent analyses based on this visual classification of individual movement tactics
186 generated results that were similar to those based on the classification using Borger and
187 Fryxell's (2012) method (not shown).

188 Subsequently, for each migratory individual, we segmented the NSD using Lavielle's
189 method (Lavielle 1999), which detects change points in a time series, to identify the dates of
190 departure and return from and to the winter range (if any) and to define individual-based
191 seasonal ranges. Dates of departure from the winter range ranged from the 6th of March to the
192 3rd of August (median = 4th of May, sd = 32 days), while return ranged from the 7th of May
193 to the 28th of October (median = 10th of September, sd = 42 days). We then used the median
194 departure and return dates across all migratory individuals to establish equivalent seasonal
195 phases for resident individuals. As a result, we subsequently analysed movement behaviour of
196 all deer during the winter period (prior to departure) and the summer period (after departure
197 and prior to return), excluding the three days prior to and following departure and return to
198 avoid the transience phase. We did not analyse data from the post-return period, (the second
199 winter), since monitoring was too short to characterise individual movement during this
200 period.

201 *Detecting sub-seasonal functional home ranges*

202 We then tested the assumption that roe deer were truly sedentary within the above defined
203 seasonal periods or, whether their seasonal ranges were composed of several sub-seasonal
204 functional home ranges (sfHR, Benhamou 2014). To do so, we segmented each individual's
205 movement path (i.e. the temporal sequence of locations) for each seasonal period using
206 Lavielle's (1999) method on the mean NSD to identify fine scale stationary states. We
207 arbitrarily considered 14 locations (i.e. 7 days) as the minimum number required to describe a
208 stationary state. We retained the most parsimonious number of segments comprising each
209 seasonal range for each individual.

210 We generated a non-linear mixed model describing the use of more than one sfHR within
211 a given season. This model was based on the mean NSD per stationary state (i.e. the number
212 of segments defined by Lavielle's method, above) for each seasonal period (Supplementary

213 materials, Appendix S1, Eq.1). To determine whether individuals were truly sedentary, we
214 compared this model with the two models of range residency developed by Börger and
215 Fryxell (2012) used above, adapted for each seasonal period (Supplementary materials,
216 Appendix S1, Eq. 2 & 3)., and retained the model with the largest concordance correlation
217 (CC) coefficient (See Supplementary materials, Appendix S2).

218 *Describing site-specific environmental conditions*

219 In order to explore the influence of environmental conditions on the propensity of roe deer to
220 use a number of sfHR within a given season, we described forage resources at each site in
221 terms of spatial heterogeneity and temporal predictability for both the winter and summer
222 periods. To identify the spatial limits of each site, we used the 100% minimum convex
223 polygon of the locations of all GPS monitored individuals at that site to which we added a
224 buffer of 1 kilometre.

225 To quantify broad scale site-specific heterogeneity in resource distribution, we used
226 the normalized difference vegetation index (NDVI) which is widely considered to be a
227 reliable index of plant productivity in general, and a good proxy of spatio-temporal variation
228 in resource availability for herbivores in particular (Hamel et al. 2009; Pettorelli et al. 2011;
229 Borowik et al. 2013). We obtained weekly values for the NDVI (pixel size = ca. 250m x
230 250m) from the University of Natural Resources and Life Sciences, Vienna (BOKU) using
231 near real-time filtered products (Whittaker smoother, (Vuolo et al. 2012; Klisch and
232 Atzberger 2016)) of the NDVI time series generated by the NASA Moderate Resolution
233 Imaging Spectroradiometer (MODIS). We used the REFMID values, supplied by BOKU,
234 which are the most stable extrapolated values of NDVI ([http://ivfl-](http://ivfl-info.boku.ac.at/index.php/eo-data-processing/real-time-modis-data-eu-only)
235 [info.boku.ac.at/index.php/eo-data-processing/real-time-modis-data-eu-only](http://ivfl-info.boku.ac.at/index.php/eo-data-processing/real-time-modis-data-eu-only) for more details).

236 To describe spatial heterogeneity in resource distribution, we measured the spatial
237 heterogeneity of the NDVI at the study site level using the standard deviation of weekly

238 values of the NDVI over all pixels of each study site (Coops et al. 1998; Coops and Culvenor
239 2000). Then, we averaged these values across weeks for each period to obtain a proxy of
240 spatial heterogeneity per seasonal period, defined above, and per year (from 2003 to 2014) for
241 each study site. Spatial heterogeneity in NDVI ranged from 0.03 (low heterogeneity) to 0.23
242 (high heterogeneity).

243 To index temporal predictability in resource distribution, we calculated temporal
244 constancy of the NDVI values across years (i.e. from 2003 to 2014) for each period at a given
245 site following the approach proposed by Colwell (1974). Constancy is a component of
246 temporal predictability (Colwell 1974) that ranges from 0 (no temporal predictability or high
247 temporal variability) to 1 (perfect temporal predictability or high temporal stability). We
248 calculated a value of temporal constancy for each period at each study site across weeks and
249 years (using weekly average values of NDVI across all pixels of a given study site). For this,
250 we built a matrix with NDVI values sorted into 10 equal 0.1 interval classes between 0 and 1
251 in rows, as indicated by English et al. (2012), and the number of the weeks in columns
252 (irrespective of the year). We counted the number of years for which we observed each class
253 of NDVI for each week, and then we calculated constancy as follows:

$$254 \quad \text{Constancy} = 1 + \frac{\sum(\sum(\text{NDVI})_{ij} / \sum\sum(\text{NDVI})_{ij} * \log(\sum(\text{NDVI})_{ij} / \sum\sum(\text{NDVI})_{ij}))}{\log(s)}$$

255 where i represents the NDVI class, j is the week of the NDVI measure, irrespective of the
256 year, and s is the number of NDVI classes.

257 As resource heterogeneity was negatively correlated with resource predictability during both
258 seasonal periods ($r = -0.50$, $p < 0.001$ during winter, $r = -0.88$, $p < 0.001$ during summer), we
259 used the residuals of the linear regression between resource predictability and resource
260 heterogeneity to index temporal predictability in resource distribution. Hence, positive values
261 indicate that temporal predictability is higher than expected for a given level of resource
262 heterogeneity.

263 *Statistical analyses*

264 To analyse the link between space use behaviour of roe deer with spatial heterogeneity and
265 temporal predictability in resource distribution, we exploited the extremely marked variation
266 among study sites in environmental conditions, performing the analyses at the population
267 level. We were unable to analyse within-population (i.e. individual level) variation in these
268 relationships due to the difficulty in measuring the distribution of resources within each study
269 site at a sufficiently fine-scale. To describe space use behaviour, we used two individual-
270 based metrics. First, we calculated the number of sfHR used by each individual during each
271 period to describe the degree to which an individual was truly sedentary within a given
272 seasonal range. Second, we calculated the median distance between the centres of all pairs of
273 sfHR for each individual and each period to index the degree of spatial separation among
274 functional ranges (sfHR distance). We log-transformed this quantity to achieve normality.

275 First, we used cumulative link mixed models, which are appropriate for ordinal data
276 (clmm, Christensen 2015a), to analyse variation in the number of sfHR within an individual's
277 seasonal range (wintering or breeding period) in relation to annual movement tactic (resident
278 vs. migratory), sex (male vs. female, because males are seasonally territorial in this species,
279 Vanpé *et al.*, 2009) and two continuous descriptors of site-specific resource distribution:
280 spatial heterogeneity and temporal predictability. We included the individual's period-specific
281 log-transformed median sfHR size as an additive factor to control for variation in absolute
282 resource availability among individuals (Morellet *et al.* 2013). For that, we calculated sfHR
283 size using the fixed kernel method with the 90% isopleths. Thus, the most complex model
284 contained two three-way interactions among movement tactic, sex and resource heterogeneity
285 and among movement tactic, sex and resource predictability, plus the logarithm of the median
286 sfHR size as a fixed effect, and the study site as a random effect to control for repeated
287 measures (individuals) per population. As preliminary analyses showed that all individuals

288 occupied at least two sfHR per season, but that most individuals used less than six, we
289 analysed the number of sfHR as a categorical variable with five modalities from two to six or
290 more (Supplementary material, Appendix S6). Note, to simplify interpretation, we only
291 present results contrasting the probabilities of occupying '2' or '6 or more' sfHR (we present
292 complete results in Supplementary material, Appendix S7 & S8).

293 Second, we used linear mixed-effects models (lmer) to investigate variation in log-
294 transformed inter-sfHR distance, including period, movement tactic, sex and resource
295 heterogeneity and resource predictability for each study site. We built the equivalent set of
296 models described above for analysing variation in inter-sfHR distance.

297 For model selection, we used Akaike's Information Criterion (AIC, Burnham and
298 Anderson 2002) and the number of parameters to select the most parsimonious model that
299 best described the data. All analyses were performed in R version 3.1.2 (R Development Core
300 Team 2014).

301 **Results**

302 *Roe deer use sub-seasonal functional home ranges: the multi-range tactic*

303 All individuals occupied more than one sfHR during a given season. Indeed, the model based
304 on the occupation of more than one sfHR received more support than the two residency
305 models (the CC value was highest for the multi-range model in all 572 cases, corresponding
306 to 286 individuals for each period). Individuals occupied a sfHR for at least 7 days (by
307 definition, see Methods), and up to 186 days, during the wintering period (median = 10 days),
308 and for at least 7 days, and up to 223 days, during the breeding period (median = 24 days).

309 *How many sub-seasonal functional home ranges?*

310 The most parsimonious model describing variation in the number of sfHR (from two to six or
311 more) during the winter period included movement tactic and the log-transformed sfHR size
312 ($\Delta AIC = 0.0$, AIC weight = 0.413, $df = 7$) (Supplementary materials, Appendix S3 & S5).

313 During the winter period, the results were partially in accordance with our expectations. First,
314 in accordance with our prediction, migratory individuals occupied fewer sfHR than resident
315 individuals: for migrants, the probability of occupying '2' sfHR was 331% higher than the
316 probability of occupying '6 or more' sfHR, while for resident individuals the probability of
317 occupying '2' sfHR was 29% lower than the probability of occupying '6 or more' sfHR (Fig.
318 2a). Second, the number of sfHR that individuals used decreased as the median size of the
319 sfHR they occupied increased. Indeed, the probability of occupying '6 or more' sfHR was
320 92% lower than the probability of occupying '2' sfHR for an individual occupying a large
321 (sfHR size = 410 ha) sfHR, whereas it was 771% higher for an individual occupying a small
322 (sfHR size = 22 ha) sfHR (Fig. 2b).

323 The most parsimonious model describing variation in the number of sfHR (from two
324 to six or more) during the summer period included sex and the log-transformed sfHR size
325 ($\Delta AIC = 0.92$, AIC weight = 0.148, $df = 7$) (Supplementary materials, Appendix S3 & S5).
326 During the summer period, both sexes had a much higher probability of using 2 sfHR than '6
327 or more' sfHR, but this difference was slightly more marked among females (the probability
328 of using '6 or more' sfHR was 98% lower than the probability of using 2 sfHR) than among
329 males (the probability of using '6 or more' sfHR was 96% lower than the probability of using
330 2 sfHR) (Fig. 3a). Finally, as above, the number of sfHR that individuals used decreased as
331 the median size of the sfHR they occupied increased. Indeed, the probability of occupying '6
332 or more' sfHR was 99% lower than the probability of occupying '2' sfHR for an individual
333 occupying a large (sfHR size = 450 ha) sfHR, whereas it was 620% higher for an individual
334 occupying a small (sfHR size = 20 ha) sfHR (Fig. 3b).

335

336 *Spatial separation among sub-seasonal functional home ranges*

337 The most parsimonious model describing variation in the log-transformed distance among
338 sfHR during the wintering period included the two two-way interactions between resource
339 heterogeneity and movement tactic and between resource heterogeneity and sex, with the
340 additive effect of the log-transformed sfHR size ($\Delta AIC = 0.6$, AIC weight = 0.274, $df = 9$)
341 (Supplementary materials, Appendix S4 & S5). First, in accordance with our prediction,
342 spatial separation between pairs of sfHR during the wintering period increased with
343 increasing resource heterogeneity, except among resident females (Fig. 4a and b). Indeed,
344 when resource heterogeneity was high (resource heterogeneity = 0.26), the log-transformed
345 inter-sfHR distance was, on average, 13% (5% for resident males and 27% for migratory
346 males) higher than for an individual living in the least heterogeneous study site (resource
347 heterogeneity = 0.08). However, for resident females, this distance was 13% lower when
348 resource heterogeneity was high. Finally, spatial separation among sfHR during the winter
349 period was 87% higher for an individual which occupied the largest sfHR (median sfHR size
350 = 410 ha) compared to an individual occupying the smallest sfHR (median sfHR size = 22 ha)
351 (Fig. 4c).

352 The best model describing variation in the log-transformed inter-sfHR distance during
353 the summer period included resource predictability in addition to sex and the median log-
354 transformed sfHR size ($\Delta AIC = 1.86$, AIC weight = 0.143, $df = 6$) (Supplementary materials,
355 Appendix S4 & S5). First, contrary to our prediction, the distance between sfHR was 12%
356 higher in the most predictable study site compared to the most unpredictable one (Fig. 5a).
357 Secondly, the distance between sfHR was 10% higher for females than for males (Fig. 5b).
358 Finally, the spatial separation among sfHR during the summer period was 94% higher for an
359 individual inhabiting the largest sfHR (median sfHR size = 450 ha) compared to an individual
360 inhabiting the smallest sfHR (median sfHR size = 20 ha) (Fig. 5c).

361 **Discussion**

362 Animals are considered sedentary when their routine movements are centred on revisited
363 areas (Papi 1992), leading to the emergence of a stable home range which may be occupied
364 for a season, or for several years (Börger et al. 2008). Sedentary behaviour is a defining
365 feature of the resident movement tactic, however, migratory animals may also be seasonally
366 sedentary within each of their distinct seasonal ranges (Börger et al. 2008; Mueller and Fagan
367 2008; Van Moorter et al. 2009). Here, we focused on movements at the within-season scale,
368 when many large herbivores (Börger et al. 2008) and, in particular, roe deer (Hewison et al.
369 1998), are presumed to be sedentary, whether or not they migrate. Based on a comprehensive
370 analysis of movement behaviour of 15 populations across Europe, we demonstrated that roe
371 deer are never truly sedentary at the seasonal scale. Instead, within any given season, both
372 migratory individuals and residents occupied at least two (and up to nine) spatially distinct
373 sfHR. We suggest that this constitutes an overlooked movement mode, the multi-range tactic,
374 which allows large herbivores to track spatio-temporal variation in the distribution of
375 available resources and, thereby, to cope with changing environmental conditions. Indeed, we
376 found that this space-use behaviour varied in relation to variation in environmental conditions
377 across the European continent.

378 At the annual scale, we were able to assign a given individual to either a residency or a
379 migratory tactic. However, when we analysed space use behaviour at the finer within season
380 temporal scale, we found that all roe deer in this study, covering widely contrasting
381 environmental conditions, adopted the multi-range tactic during a given season. A similar
382 pattern of space use behaviour has been documented in two African large herbivores, the
383 sable antelope, the African savannah buffalo (Owen-Smith et al. 2010; Cornélis et al. 2011;
384 Benhamou 2014). These authors showed that the ranges of these animals were composed of
385 several distinct areas which were exploited for several days or weeks. Thus, the use of
386 multiple sfHR to track available food resources seems to be potentially widespread among

387 large herbivores. Here, to understand the proximal drivers of this seasonal space use
388 behaviour, we explored how spatial heterogeneity and temporal predictability of resource
389 distribution influenced variation in the number and the spatial distribution of the sub-seasonal
390 ranges that an individual exploits.

391 Mueller et al. (2011), focusing on movements at the annual scale, documented longer
392 seasonal migrations among species inhabiting areas where primary productivity was spatially
393 heterogeneous at a large-scale, but shorter migrations for species inhabiting environments
394 with relatively low spatial heterogeneity. In an analogous fashion, but at the seasonal scale,
395 we expected individuals to be less sedentary in heterogeneous environments than in
396 homogeneous environments, occupying a higher number of more spatially distant sfHR. Our
397 analyses only partially supported this prediction as, in heterogeneous study areas, the distance
398 between sfHR was indeed higher compared to homogeneous sites, but only during winter.
399 Furthermore, the number of sfHR that deer occupied did not vary in relation to resource
400 heterogeneity. One explanation for this discrepancy could be linked to the relatively coarse
401 spatial resolution of the NDVI metric that limits our ability to quantify small-scale variations
402 in resource distribution at a level that is informative for individuals. Indeed, Van Moorter et
403 al. (2013) have shown that movements at a particular scale are driven by changes in the net
404 profitability of trophic resources at the corresponding scale. As a result, a finer-scale measure
405 of spatial heterogeneity could help us to better understand why roe deer use a number of
406 spatially distinct sub-seasonal functional home ranges, even in apparently homogeneous
407 habitats.

408 Because nomadism is considered to be a response to unpredictability (Mueller and
409 Fagan 2008), we expected roe deer to be less sedentary in unpredictable environments than in
410 predictable environments. We found little support for this prediction. Our results indicated
411 that, while controlling for resource heterogeneity, resource predictability had no effect on

412 variation in the number of sfHR, and only a very weak effect on the distance among sfHR,
413 and this only during summer. During this period, roe deer adopted a space use behaviour that
414 was somewhat more similar to a nomadic tactic when in more predictable environments,
415 moving further among sub-seasonal functional ranges than in unpredictable environments.
416 Given the lack of support for our predictions, we suggest that other factors such as predation
417 risk or human disturbance could drive roe deer to switch periodically from one locality to
418 another. Theory predicts that prey should perform frequent random movements to minimise
419 the probability of encountering predators (Mitchell and Lima 2002). Indeed, previous studies
420 have shown that herbivores modify their habitat selection or switch location after an
421 encounter with a predator (Latombe et al. 2014; Lone et al. 2016).

422 From the point of view of landscape complementation (Dunning et al. 1992), an
423 animal's home range must necessarily contain a combination of all non substitutable
424 resources required for survival and reproduction. Indeed, variation in home range size has
425 previously been shown to reflect variation in resource availability, integrating interactions
426 among local weather, climate and environmental seasonality (Morellet et al. 2013). Hence, all
427 things being equal, animals will occupy larger home ranges when resources are more sparsely
428 distributed in space (Saïd et al. 2005; Van Beest et al. 2011). Our analyses demonstrated
429 strong relationships between sub-seasonal range size with both the number of sfRH and the
430 distance between them. These results indicate that when resources are abundant, deer
431 sequentially exploit a number of short-term functional ranges that are within close proximity.
432 In contrast, when resources are limiting, deer tend to relocate less frequently, possibly due to
433 the costs of doing so in terms of predation risk, mortality or energy (Hein et al. 2012;
434 Johansson et al. 2014), but move greater distances when they do so to locate a new functional
435 sub-seasonal range. This individual variation in space use is further demonstration of the
436 extensive behavioural plasticity that roe deer express in relation to prevailing environmental

437 conditions (Cagnacci et al. 2011; Morellet et al. 2013; Lone et al. 2016) which has driven the
438 undoubted recent success story of this species across Europe (Linnell et al. 1998).

439 Finally, as expected, migratory individuals occupied fewer sfHR than resident
440 individuals. Seasonal migration is a tactic designed to cope with seasonal changes in the
441 spatial distribution of resources (Fryxell and Sinclair 1988; Mueller and Fagan 2008). As a
442 result, it seems that migratory individuals do not have to shift their ranges within seasons, as
443 they are able to adjust their movements to spatial variation in the distribution of available
444 forage at the between-season scale.

445 In conclusion, we suggest that the multi-range tactic is an individual-level behavioural
446 response to cope with spatio-temporal variation in the distribution of resources when this
447 occurs at the within seasonal scale. This tactic appears to be ubiquitous in the roe deer,
448 occurring across its entire European distribution, and encompassing a wide gradient of
449 environmental conditions. We suggest that large herbivores may adopt this tactic when
450 environmental conditions fluctuate spatially and temporally, independently of seasonal
451 variations. In the present context of climate change, predictions of more frequent and intense
452 climatic events (IPCC 2014) may mean that an increasing number of large herbivore
453 populations adopt the multi-range tactic, combining more frequent spasmodic movements
454 interspersed with short periods of sedentarism, to track available resources.

455

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467

468 **Ethical approval:** All applicable institutional and/or national guidelines for the care and
469 use of animals were followed.

470

471 **Data accessibility**

472 Data used for this study are accessible on EURODEER website (<http://eurodeer.org>) on
473 demand.

474

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610

Tables

Table 1: Number of males, females, migratory individuals and residents per study site.

Study site id	Study site name	Migrants	Residents	Females	Males
a	Aspromonte - Italy	4	2	4	2
b	Aurignac - France	20	86	66	40
c	Baden-Wurttemberg - Germany	5	19	22	2
d	Bavarian Forest National Park - Germany	28	35	36	27
e	Šumava National Park – Czech republic	0	2	1	1
f	Bernese Oberland - Switzerland	9	15	19	5
g	Bialowieza National Park - Poland	2	5	7	0
h	Brandenburg - Germany	2	0	2	0
i	Grimso - Sweden	1	0	0	1
j	Koberg - Sweden	5	9	10	4
k	Leoben - Austria	4	4	5	3
l	Matese - Italy	0	3	3	0
m	Monte Bondone - Italy	4	4	4	4
n	Rendena Giudicarie – Italy	6	3	6	3
o	NINA west - Norway	3	6	5	4
Total		93	193	190	96

Figure legends

Figure 1: Location of study sites where roe deer were monitored using GPS collars. Letters correspond to the ‘study site id’ in Table 1.

Figure 2: Graphical representation of the best model describing variation in the number of sub-seasonal home ranges (sfHR) during the winter period, represented here by the probability of occupying either ‘2’ or ‘6 or more’ sfHR (for simplification, intermediate values of sfHR are excluded from the graph, see Supplementary material, Appendix S7), as a function of movement tactic (a) and the log-transformed sfHR size (b). Dotted lines represent the 95% confidence intervals resulting from the model predictions. We fixed the log-transformed sfHR size at the mean and movement tactic as residency when not represented in the following figures.

Figure 3: Graphical representation of the best model describing variation in the number of sub-seasonal home ranges (sfHR) during the summer period, represented here by the probability of occupying either ‘2’ or ‘6 or more’ sfHR (for simplification, intermediate values of sfHR are excluded from the graph, see Supplementary material, Appendix S8), as a function of sex (a) and the log-transformed ssHR size (b). Dotted lines represent the 95% confidence intervals resulting from the model predictions. We fixed the log-transformed sfHR size at the mean and sex as female when not represented in the following figures.

Figure 4: Graphical representation of the best model describing variation in the log-transformed distance among sub-seasonal home ranges (sfHR) during the winter period, as a function of the two-way interaction between resource heterogeneity and movement tactic and between resource heterogeneity and sex, represented as a three-way interaction (a: females, b: males), and the log-transformed median sfHR size (c). In order to better visualize the raw data, we plotted the mean log-transformed sfHR distance for each study site. Dotted lines represent the 95% confidence intervals resulting from the model predictions, error bars

represent the standard deviation of log-transformed inter-sfHR distance, and the letters correspond to the ‘study site id’ in Table 1. We fixed resource heterogeneity and the log-transformed sfHR size at the mean, movement tactic as residency and sex as female when not represented in the following figures.

Figure 5: Graphical representation of the best model describing variation in the log-transformed distance among sub-seasonal home ranges (sfHR) during the summer period, as a function of resource predictability (a), sex (b) and the log-transformed median sfHR size (c). In order to better visualize the raw data, we plotted the mean log-transformed sfHR distance for each study site. Dotted lines represent the 95% confidence intervals resulting from the model predictions, error bars represent the standard deviation of log-transformed inter-sfHR distance, and the letters correspond to the ‘study site id’ in Table 1. We fixed resource predictability and the log-transformed sfHR size at the mean and sex as female when not represented in the following figures.

Fig. 1

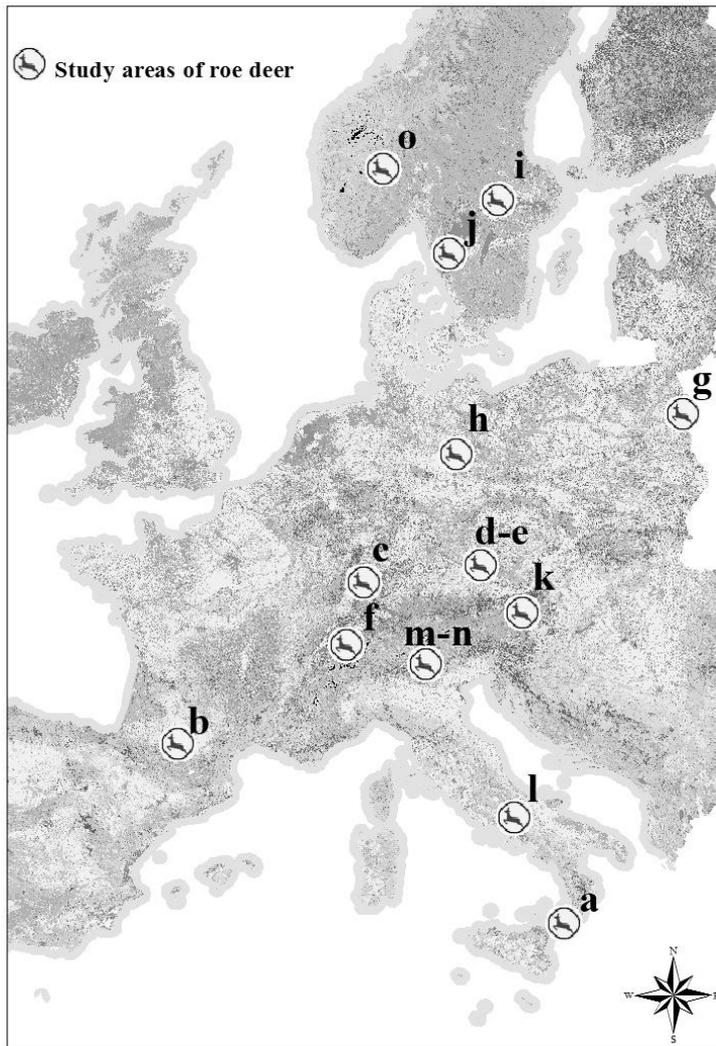


Fig. 2

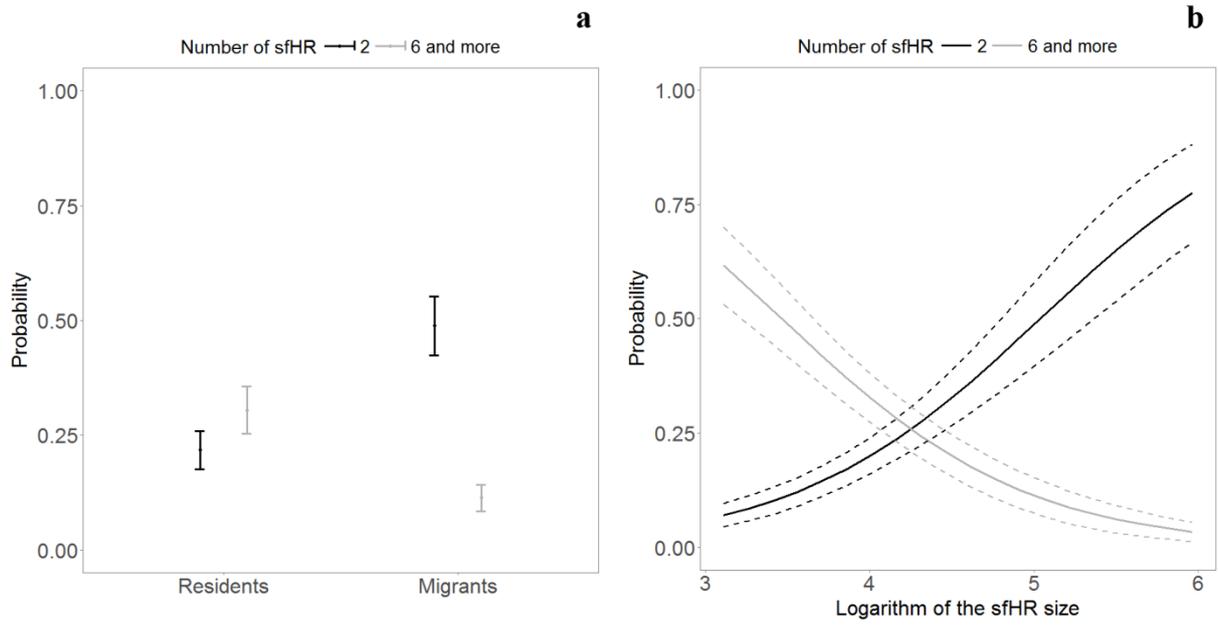


Fig. 3

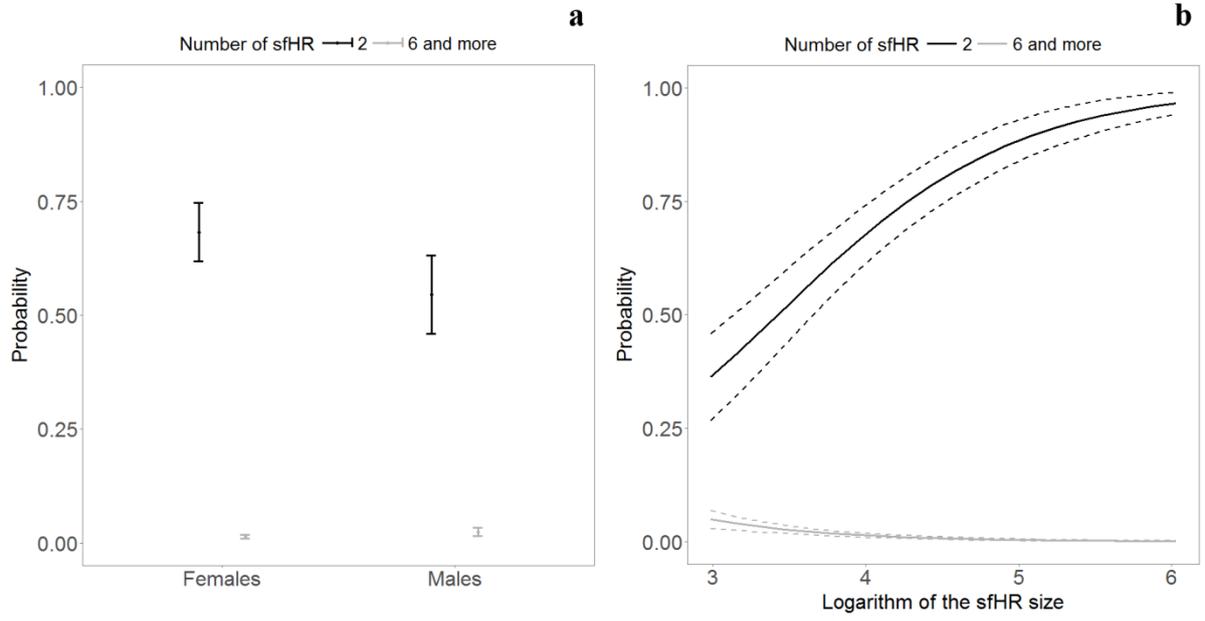


Fig. 4

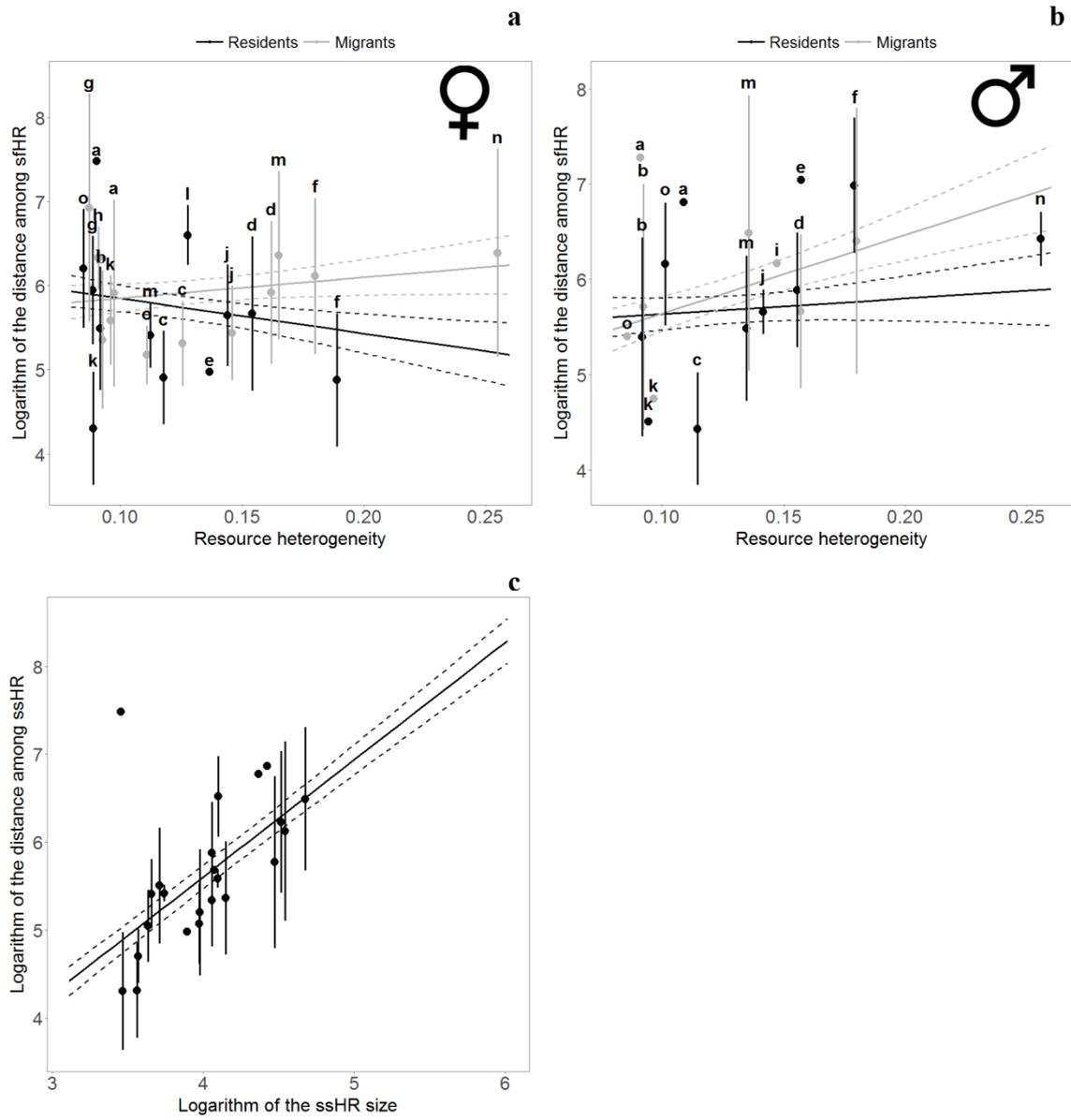
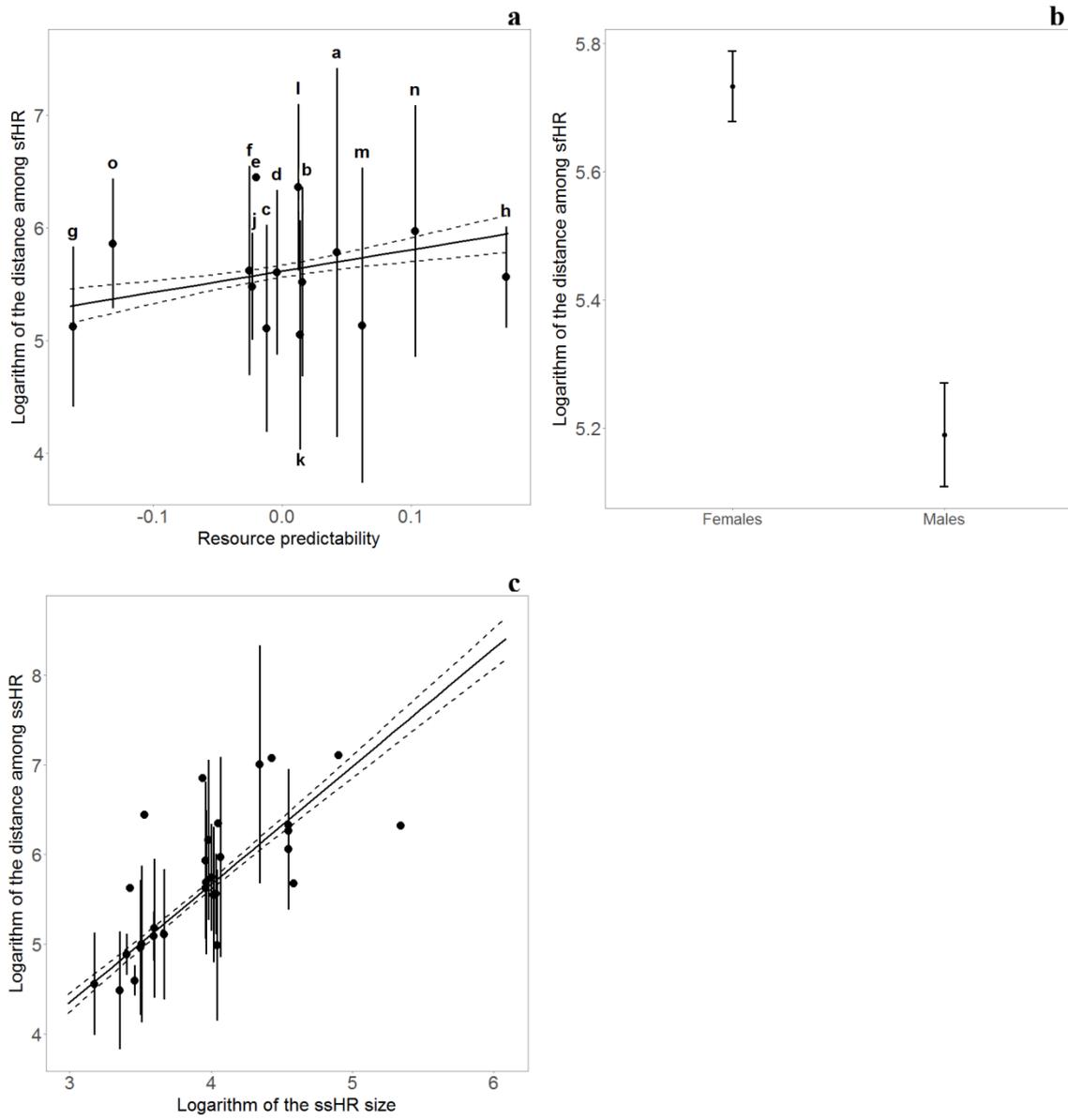


Fig. 5





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