

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

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56 Abstract

Much research on large herbivore movement has focused on the annual scale to distinguish 57 between resident and migratory tactics, commonly assuming that individuals are sedentary at 58 the within-season scale. However, apparently sedentary animals may occupy a number of sub-59 seasonal functional home ranges (sfHR), particularly when the environment is spatially 60 heterogeneous and/or temporally unpredictable. The roe deer (Capreolus capreolus) 61 62 experiences sharply contrasting environmental conditions due to its widespread distribution, but appears markedly sedentary over much of its range. Using GPS monitoring from 15 63 populations across Europe, we evaluated the propensity of this large herbivore to be truly 64 sedentary at the seasonal scale in relation to variation in environmental conditions. We 65 studied movement using net square displacement to identify the possible use of sfHR. We 66 67 expected that roe deer should be less sedentary within seasons in heterogeneous and unpredictable environments, while migratory individuals should be seasonally more sedentary 68 than residents. Our analyses revealed that, across the 15 populations, all individuals adopted a 69 70 multi-range tactic, occupying between two and nine sfHR during a given season. In addition, we showed that (i) the number of sfHR was only marginally influenced by variation in 71 resource distribution, but decreased with increasing sfHR size; and (ii) the distance between 72 73 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 74 75 among large herbivores, allowing animals to track spatio-temporal variation in resource distribution and, thereby, to cope with changes in their local environment. 76

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78 Key words: migration, residency, sub-seasonal functional home range, plasticity, roe deer

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80 Introduction

Movement is a fundamental characteristic of life which influences the survival and 81 reproduction of organisms and, more generally, individual fitness and population dynamics 82 (Turchin 1991; Revilla and Wiegand 2008). Following the Marginal Value Theorem 83 (Charnov 1976), while foraging, an individual moves within a patch (sensu Wiens 1976), 84 searching intensively for food, before leaving to search more widely for a new patch when the 85 86 energetic benefit of the first patch has decreased below the average value of the alternative patches (Charnov 1976; Van Moorter et al. 2009). In slowly regenerating environments, 87 animals should engage in thorough exploration to locate suitable patches. However, in rapidly 88 regenerating environments, individuals are assumed to memorise the value and location of a 89 given patch (Bracis et al. 2015; Riotte-Lambert et al. 2015), returning to previously visited 90 91 patches periodically. This process leads to the emergence of a stable home range to which individuals restrict their movements in order to maximise resource acquisition (Brown and 92 Orians 1970; Riotte-Lambert et al. 2015). Therefore, individuals of many species appear to be 93 94 sedentary at this spatio-temporal scale, occupying a stable home range over a long time span 95 (season and year). Site fidelity is widespread in the animal kingdom and has fundamental 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 101 102 resources (Nathan et al. 2008) which encompass changes in environmental space (Van 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 107 108 requirements, in which case it cannot be considered truly sedentary, even at the seasonal scale (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* 116 117 caffer brachyceros) (Owen-Smith et al. 2010; Cornélis et al. 2011; Benhamou 2014). However, no study has yet attempted to link the propensity of individuals to adopt this multi-118 119 range tactic with spatial and temporal variation in the prevailing environmental conditions.

120 Here, we used the EURODEER database (http://www.eurodeer.org, see methods) to 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 spatially 131 heterogeneous compared to homogeneous environments (Mueller and Fagan 2008; Mueller et 132 al. 2011). Analogously to the nomadic movement tactic (Mueller and Fagan 2008), we also 133 expected individuals to be less sedentary in temporally unpredictable environments, or at least 134 in environments where resources vary more markedly over time at the within-season scale, 135 compared to more predictable environments. Finally, we expected that, within a given 136 137 seasonal range, migratory animals would be more sedentary than residents since they migrate during spring and fall so that they are able to adjust their habitat use to seasonal variations in 138 food resource abundance and/or quality at this scale (Fryxell and Sinclair 1988). 139

140 Material and methods

141 Study areas and GPS data

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, 147 148 net traps or box traps depending on study site. All capture and marking procedures were done 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. 150 151 In order to standardise the data for inter-population comparisons, for each individual, we 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

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Discrimination of individual movement tactics

Of the recently published emerging tools for determining movement tactic (Gurarie et al. 156 2017; Spitz et al. 2017), we first used a common method proposed by Börger and Fryxell 157 (2012), which has been proven useful for large mammals (Bunnefeld et al. 2011; Bischof et 158 al. 2012), based on the net squared displacement (NSD), *i.e.* the Euclidian distance between 159 the starting location and all subsequent locations of an individual over time (Turchin 1998), to 160 161 determine each individual's annual movement tactic: migration, residency or dispersal. We considered two models of range residency, one with a constant NSD (the mean), and one with 162 163 a linear increase of NSD before reaching an asymptote; we considered one model of migration including approximate dates of departure and return between seasonal ranges, and a 164

model of dispersal with an approximate date of departure (see Bunnefeld et al. 2011 and 165 Börger and Fryxell 2012 for more details on these models). In order to identify which of these 166 models best described the movement behaviour of a given individual, we used the system of 167 168 non-linear mixed models proposed by Börger and Fryxell (2012) which links theoretical expectations to movement data. For model selection, as recommended by Börger and Fryxell 169 (2012), we retained the model with the largest concordance correlation (CC), expressing the 170 goodness of fit for each model (Huang et al. 2009). Because the assigned movement tactic 171 using this method did not always closely fit the data, we also visually examined the NSD 172 trajectories to determine each individual's annual movement tactic by eye (Bischof et al. 173 2012). We based our visual classification on the patterns of NSD typically observed for 174 migratory individuals, residents and dispersers, following Börger and Fryxell (2012). That is, 175 we assumed that the individual was resident when the NSD was relatively constant, or 176 177 initially increased linearly before rapidly reaching an asymptote. We assumed that the individual was migratory when the NSD was constant before increasing rapidly during spring 178 179 to reach a plateau during summer, then decreased during fall, returning to its initial value. 180 Finally, we assumed that the individual had dispersed when the NSD was constant before increasing rapidly to reach a plateau with no further increase or decrease. We then verified 181 that individuals which were classified as dispersers did not return to their point of departure 182 during subsequent monitoring, after the 15th of November. If they did (53 of 65 animals 183 originally classified as dispersers), these individuals were considered as migratory. We 184 excluded the remaining dispersers (N = 12) from subsequent analyses as movement patterns 185 during dispersal are governed by different ultimate causes than those involved in range 186 residency or migration (Bowler and Benton 2005; Chapman et al. 2014). After visual 187 reclassification, our data set included 193 residents and 93 migratory individuals. Note that 188 subsequent analyses based on this visual classification of individual movement tactics 189

190 generated results that were similar to those based on the classification using Borger and191 Fryxell's (2012) method (not shown).

Subsequently, for each migratory individual, we segmented the NSD using Lavielle's 192 method (Lavielle 1999), which detects change points in a time series, to identify the dates of 193 departure and return from, and to, the winter range (if any) and to define individual-based 194 seasonal ranges. Dates of departure from the winter range ranged from the 6th of March to the 195 3^{rd} of August (median = 4^{th} of May, sd = 32 days), while return ranged from the 7^{th} of May to 196 the 28^{th} of October (median = 10^{th} of September, sd = 42 days). We then used the median 197 departure and return dates across all migratory individuals to establish equivalent seasonal 198 phases for resident individuals. As a result, we subsequently analysed movement behaviour of 199 all deer during the winter period (prior to departure) and the summer period (after departure 200 and prior to return), excluding the three days prior to and following departure and return to 201 202 avoid the transience phase. We did not analyse data from the post-return period, (the second winter), since monitoring periods were too short to characterise individual movement during 203 204 this period.

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Detecting sub-seasonal functional home ranges

207 We then tested the assumption that roe deer were truly sedentary within the above defined seasonal periods or, whether their seasonal ranges were composed of several sub-seasonal 208 209 functional home ranges (sfHR, Benhamou 2014). To do so, we segmented each individual's movement path (i.e. the temporal sequence of locations) for each seasonal period using 210 Lavielle's (1999) method on the mean NSD to identify fine scale stationary states. Following 211 several studies which have examined variation of weekly home ranges in response to variation 212 in resource distribution (Rivrud et al. 2010; Van Beest et al. 2011; Morellet et al. 2013), we 213 considered 14 locations (*i.e.* 7 days) as the minimum number required to describe a stationary 214

state. Because this choice of a minimum time span is somewhat arbitrary, we repeated the segmentation approach twice using minimum values of 5 and of 9 days (*i.e.* 7 ± 2 days) to assess the sensitivity of our results to the length of the temporal window. Finally, we retained the most parsimonious number of segments comprising each seasonal range for each individual (Figs. 2 & 3). We again repeated this step when considering 5 and 9 days as the minimum temporal windows.

We generated a non-linear mixed model describing the use of more than one sfHR within 221 a given season. This model was based on the mean NSD per stationary state (*i.e.* the number 222 of segments defined by Lavielle's method) for each seasonal period (Supplementary 223 materials, Appendix 1, Eq. A1). To determine whether individuals were truly sedentary, we 224 compared this model with the two models of range residency developed by Börger and 225 Fryxell (2012) used above, adapted for each seasonal period (Supplementary materials, 226 227 Appendix 1, Eq. A2 & A3), and retained the model with the largest CC (See Supplementary materials, Appendix 1, Table A1). 228

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Describing site-specific environmental conditions

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

To quantify broad scale site-specific heterogeneity in resource distribution, we used the normalized difference vegetation index (NDVI). NDVI is widely used in ecological research and management as a proxy for biomass (Kerr and Ostrovsky 2003; Pettorelli et al.

2011) and forage quality available for ungulates (Hamel et al. 2009; Borowik et al. 2013; 240 Lendrum et al. 2014; Garroutte et al. 2016) and particularly for roe deer (Morellet et al. 2013; 241 Peters et al. 2017). NDVI was used because of its simplicity, consistent coverage of all study 242 243 areas, high temporal resolution and demonstrated utility in a variety of ecosystems. Given the linear relationship between NDVI and biomass, a given variation in NDVI should index a 244 similar relative reduction in forage biomass across study sites. Hence, it provides a useful and 245 integrative measure of variation in primary productivity which we assume to be suitable for 246 247 our purpose of comparing roe deer populations inhabiting widely contrasting environmental conditions. 248

We obtained weekly values for the NDVI (pixel size = ca. 250m x 250m) from the University of Natural Resources and Life Sciences, Vienna (BOKU) using near real-time filtered products (Whittaker smoother, (Vuolo et al. 2012; Klisch and Atzberger 2016)) of the NDVI time series generated by the NASA Moderate Resolution Imaging Spectroradiometer (MODIS). We used the REFMID values, supplied by BOKU, which are the most stable extrapolated values of NDVI (http://ivfl-info.boku.ac.at/index.php/eo-data-processing/realtime-modis-data-eu-only for more details).

To describe spatial heterogeneity in resource distribution, we measured spatial heterogeneity of the NDVI at the study site level using the standard deviation of weekly values of the NDVI over all pixels of each study site (Coops et al. 1998; Coops and Culvenor 2000). Then, we averaged these values across weeks for each period to obtain a proxy of spatial heterogeneity per seasonal period, defined above, and per year (from 2003 to 2014) for each study site. Spatial heterogeneity in NDVI ranged from 0.03 (low heterogeneity) to 0.23 (high heterogeneity).

To index temporal predictability in resource distribution, we calculated temporal constancy of the NDVI values across years (*i.e.* from 2003 to 2014) for each period at a given

site following the approach proposed by Colwell (1974). Constancy is a component of 265 temporal predictability (Colwell 1974) that ranges from 0 (no temporal predictability or high 266 temporal variability) to 1 (perfect temporal predictability or high temporal stability). We 267 calculated a value of temporal constancy for each period at each study site across weeks and 268 years (using weekly average values of NDVI across all pixels of a given study site). For this, 269 we built a matrix with NDVI values sorted into 10 equal 0.1 interval classes between 0 and 1 270 in rows, as indicated by English et al. (2012), and the number of the weeks in columns 271 (irrespective of the year). We counted the number of years for which we observed each class 272 of NDVI for each week, and then we calculated constancy as follows: 273

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$$Constancy = 1 + \frac{\sum \frac{\sum NDVI_{ij}}{\sum \sum NDVI_{ij}} \log(\frac{\sum NDVI_{ij}}{\sum \sum NDVI_{ij}})}{\log(s)}$$

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

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

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284 *Statistical analyses*

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

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How many sub-seasonal functional home ranges?

We used Generalised Linear Mixed-Effects Models (glmm), with a Poisson distribution and a 298 log-link function, to analyse variation in the number of sfHR within an individual's seasonal 299 300 range (wintering or breeding period) in relation to annual movement tactic (resident vs. migratory), sex (male vs. female, because males are seasonally territorial in this species, 301 Vanpé et al., 2009) and two continuous descriptors of site-specific resource distribution: 302 303 spatial heterogeneity and temporal predictability. We included the individual's period-specific log-transformed median sfHR size as an additive factor to control for variation in absolute 304 resource availability between individuals (Morellet et al. 2013). Thus, the most complex 305 306 model contained two three-way interactions among movement tactic, sex and resource heterogeneity and among movement tactic, sex and resource predictability, plus the logarithm 307 308 of the median sfHR size as a fixed effect, and the study site as a random effect to control for repeated measures (individuals) per population. As the number of sfHR occupied by an 309 individual ranged from 2 to 9, we subtracted 2 from the total number of sfHR per individual 310 per season in order to conform to a Poisson distribution (Haight 1967). The number of sfHR 311 then ranged from 0 to 7. We estimated sfHR size using fixed-kernel methods considering (i) 312 an ad hoc approach to select the optimal smoothing parameter h for each sfHR estimate and 313

(ii) a constant h set to the overall median of the smoothing parameters estimated by the ad hoc
approach for each individual sfHR estimate. Finally, sfHR size was estimated at both 90%
and 50% isopleths.

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Spatial separation among sub-seasonal functional home ranges

We used Linear Mixed-Effects models (lme) to investigate variation in log-transformed intersfHR distance, including period, movement tactic, sex and resource heterogeneity and resource predictability for each study site. We built the equivalent set of models described above for analysing variation in inter-sfHR distance.

For model selection, we used Akaike's Information Criterion (AIC, Burnham and Anderson 2002) and the number of parameters to select the most parsimonious model that best described the data. All analyses were performed in R version 3.1.2 (R Development Core Team 2014) using *glmer* and *lmer* functions in the *lme4* library.

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328 **Results**

Our overall results were robust with respect to the choice of method to estimate sfHR size, the isopleth used (see Supplementary materials, Appendix 1, Fig. A7 to A15), and the minimum number of days required to describe a sfHR (see Supplementary materials, Appendix 1, Fig. A1 to A6). Thus, in the following, we only present results using the 90% fixed-kernel method based on a constant h and 7 days as the minimum temporal window for a sfHR.

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Roe deer use sub-seasonal functional home ranges: the multi-range tactic All individuals occupied more than one sfHR during a given season. Indeed, the model based on the occupation of more than one sfHR received more support than the two residency models (the CC value was highest for the multi-range model in all 572 cases, corresponding to 286 individuals for the two periods, Supplementary materials, Appendix 1, Table A1).
Individuals occupied a sfHR for at least 7 days (by definition, see Methods), and up to 186
days during the wintering period (median = 10 days), and for at least 7 days, and up to 223
days, during the breeding period (median = 24 days).

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How many sub-seasonal functional home ranges?

The most parsimonious model describing variation in the number of sfHR during the winter 345 period included the two-way interaction between resource predictability and movement tactic 346 and the log-transformed sfHR size ($\Delta AIC = 0.0$, AIC weight = 0.319, df = 6) (Supplementary 347 348 materials, Appendix 1, Table A2 & A4). During the winter period, the results were partially in accordance with our expectations. Firstly, in accordance with our second prediction, resident 349 individuals occupied 1.5 fewer sfHR in the most predictable study site compared to the most 350 351 unpredictable one. But contrary to this prediction, migrants tended to occupy one more sfHR 352 in the most predictable study site compared to the most unpredictable one. However, in accordance with our third prediction, except in the most predictable study site where residents 353 and migrants occupied the same number of sfHR, a resident individual used more sfHR on 354 average than a migrant individual (Fig. 4a). Secondly, the number of sfHR that individuals 355 356 used decreased as the median size of the sfHR they occupied increased. Indeed, an individual that occupied a large (sfHR size = 410 ha) sfHR had 4 fewer sfHR than an individual that 357 occupied a small (sfHR size = 22 ha) sfHR (Fig. 4b). 358

The most parsimonious model describing variation in the number of sfHR during the summer period included sex and the log-transformed sfHR size ($\Delta AIC = 0.0$, AIC weight = 0.167, df = 4) (Supplementary materials, Appendix 1, Table A2 & A4). During the summer period, the difference in the number of sfHR occupied by males and females was low. Indeed, males used, on average, 0.4 sfHR more than females (Fig. 4c). Finally, as above, the number of sfHR that individuals used decreased as the median size of the sfHR they occupied increased. Indeed, an individual that occupied a large (sfHR size = 450 ha) sfHR had 2 sfHR less than an individual that occupied a small (sfHR size = 20 ha) sfHR (Fig. 4c).

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Spatial separation among sub-seasonal functional home ranges

The most parsimonious model describing variation in the log-transformed distance among 369 370 sfHR during the wintering period included the two two-way interactions between resource heterogeneity and movement tactic and between resource heterogeneity and sex, with the 371 additive effect of the log-transformed sfHR size ($\Delta AIC = 0.6$, AIC weight = 0.274, df = 9) 372 (Supplementary materials, Appendix 1, Table A3 & A4). First, in accordance with our first 373 prediction, spatial separation between pairs of sfHR during the wintering period increased 374 375 with increasing resource heterogeneity, except among resident females (Fig. 5a and b). Indeed, when resource heterogeneity was high (resource heterogeneity = 0.26), the distance 376 377 among sfHR of an individual was, on average, 700 meters (365 meters for resident males and 378 1050 meters for migratory males), while for an individual living in the least heterogeneous study site (resource heterogeneity = 0.08) it was, on average, 300 meters. However, for 379 resident females, this distance was 350 meters lower when resource heterogeneity was high. 380 381 Finally, spatial separation among sfHR during the winter period was 3980 meters for an individual which occupied the largest sfHR (median sfHR size = 410 ha) compared to 80 382 383 meters for an individual that occupied the smallest sfHR (median sfHR size = 22 ha) (Fig. 5c).

The best model describing variation in the log-transformed inter-sfHR distance during the summer period included resource predictability in addition to sex and the median logtransformed sfHR size ($\Delta AIC = 1.86$, AIC weight = 0.143, df = 6) (Supplementary materials, Appendix 1, Table A3 & A4). Firstly, contrary to our second prediction, the distance between sfHR was 384 meters in the most predictable study site compared to 200 meters in the most unpredictable one (Fig. 6a). Secondly, the distance between sfHR was 308 meters for females
and 177 meters for males (Fig. 6b). Finally, the spatial separation among sfHR during the
summer period was 4450 meters for an individual inhabiting the largest sfHR (median sfHR
size = 450 ha) compared to 76 meters for an individual inhabiting the smallest sfHR (median
sfHR size = 20 ha) (Fig. 6c).

394

395 **Discussion**

Animals are considered sedentary when their routine movements are centred on revisited 396 397 areas (Papi 1992), leading to the emergence of a stable home range which may be occupied for a season, or for several years (Börger et al. 2008). Sedentary behaviour is a defining 398 feature of the resident movement tactic, however, migratory animals may also be seasonally 399 sedentary within each of their distinct seasonal ranges (Börger et al. 2008; Mueller and Fagan 400 2008; Van Moorter et al. 2009). Here, we focused on movements at the within-season scale, 401 402 when many large herbivores (Börger et al. 2008) and, in particular, roe deer (Hewison et al. 1998), are presumed to be sedentary. Based on a comprehensive analysis of movement 403 404 behaviour of 15 populations across Europe, we demonstrated that roe deer are never truly 405 sedentary at the seasonal scale. Instead, within any given season, both migratory individuals and residents occupied at least two (and up to nine) spatially distinct sfHR. We suggest that 406 this constitutes an overlooked movement mode, the multi-range tactic, which allows large 407 408 herbivores to track fine-scaled spatio-temporal variation in the distribution of available resources and, thereby, to cope with changing environmental conditions. Indeed, we found 409 410 that this space-use behaviour varied in relation to variation in environmental conditions across the European continent. 411

412 At the annual scale, we were able to assign a given individual to either a residency or a 413 migratory tactic. However, when we analysed space use behaviour at the finer temporal scale

within seasons, we found that all roe deer in this study adopted the multi-range tactic during a 414 415 given season. A similar pattern of space use behaviour has been documented in two African large herbivores, the sable antelope and the African savannah buffalo (Owen-Smith et al. 416 2010; Cornélis et al. 2011; Benhamou 2014). These authors showed that the ranges of these 417 animals were composed of several distinct areas which were exploited for several days or 418 weeks. Thus, the use of multiple sfHR to track available food resources seems to be 419 potentially widespread among large herbivores. Here, to understand the proximal drivers of 420 this seasonal space use behaviour, we explored how spatial heterogeneity and temporal 421 predictability of resource distribution influenced variation in the number and the spatial 422 423 distribution of the sub-seasonal ranges that an individual exploits.

Mueller et al. (2011), focusing on movements at the annual scale, documented longer 424 seasonal migrations among species inhabiting areas where large-scale primary productivity 425 426 was spatially heterogeneous, but shorter migrations for species inhabiting environments with relatively low spatial heterogeneity. Similarly, but at the seasonal scale, we expected 427 428 individuals to be less sedentary in heterogeneous environments than in homogeneous 429 environments, occupying a higher number of more spatially distant sfHR. Our analyses only partially supported this prediction as, in heterogeneous study areas, the distance between 430 sfHR was indeed higher compared to homogeneous sites, but only during winter. 431 Furthermore, the number of sfHR that deer occupied did not vary in relation to resource 432 heterogeneity. One explanation for this discrepancy could be linked to the relatively coarse 433 spatial resolution of the NDVI metric that limits our ability to quantify small-scale variations 434 in resource distribution at a level that is informative for individual animals. Indeed, Van 435 Moorter et al. (2013) have shown that movements at a particular scale are driven by changes 436 437 in the net profitability of trophic resources at the corresponding scale. As a result, a finer-scale measure of spatial heterogeneity could help us to better understand why roe deer use a number 438

439 of spatially distinct sub-seasonal functional home ranges, even in apparently homogeneous440 habitats.

Because nomadism is considered to be a response to unpredictability (Mueller and 441 Fagan 2008), we expected roe deer to be less sedentary in unpredictable environments than in 442 predictable environments. We found little support for this prediction. Our results indicated 443 that, while controlling for resource heterogeneity, resource predictability had only a weak 444 effect on the number of sfHR and the distance among sfHR. During summer, roe deer moved 445 slightly further when in predictable environments than in unpredictable ones. During winter, 446 resident roe deer adopted a space use behaviour that was somewhat similar to a nomadic tactic 447 when in more unpredictable environments, occupying slightly more sfHR than in predictable 448 environments. Migratory individuals, on the contrary, tended to occupy slightly less sfHR in 449 predictable environments than in unpredictable ones. However, these relationships were weak, 450 451 with the main difference being that resident individuals occupied more sfHR than migrants during winter. Seasonal migration is a tactic designed to cope with seasonal changes in the 452 spatial distribution of resources (Fryxell and Sinclair 1988; Mueller and Fagan 2008). As a 453 result, it seems that migratory individuals do not have to shift their ranges at the finer within-454 season scale, as they are able to adjust their movements to spatial variation in the distribution 455 456 of available forage at the larger between-season scale. Given the lack of support for our predictions, we suggest that other factors such as fine-scale variation in forage plant 457 distribution, predation risk, or human disturbance could drive roe deer to switch periodically 458 from one locality to another. Theory also predicts that prey should perform frequent random 459 movements to minimise the probability of encountering predators (Mitchell and Lima 2002). 460 Indeed, previous studies have shown that herbivores modify their habitat selection or switch 461 location after an encounter with a predator (Latombe et al. 2014; Lone et al. 2016). 462

From the point of view of landscape complementation (Dunning et al. 1992), an 463 animal's home range must necessarily contain a combination of all non-substitutable 464 resources required for survival and reproduction. Indeed, variation in home range size has 465 previously been shown to reflect variation in resource availability, integrating interactions 466 among local weather, climate and environmental seasonality (Morellet et al. 2013). Hence, all 467 things being equal, animals will occupy larger home ranges when resources are more sparsely 468 distributed in space (Saïd et al. 2005; Van Beest et al. 2011). Our analyses demonstrated 469 470 strong relationships between sub-seasonal range size with both the number of sfRH and the distance between them. These results indicate that when resources are abundant, deer 471 sequentially exploit a number of short-term functional ranges that are within close proximity. 472 In contrast, when resources are limiting, deer tend to relocate less frequently, possibly due to 473 the costs of doing so in terms of predation risk, mortality or energy (Hein et al. 2012; 474 475 Johansson et al. 2014), but move greater distances when they do so to locate a new functional sub-seasonal range. This individual variation in space use is further demonstration of the 476 extensive behavioural plasticity that roe deer express in relation to prevailing environmental 477 conditions (Cagnacci et al. 2011; Morellet et al. 2013; Lone et al. 2016) which has driven the 478 479 undoubted recent success story of this species across Europe (Linnell et al. 1998).

480 In conclusion, we suggest that the multi-range tactic is an individual-level behavioural response to cope with spatio-temporal variation in the distribution of resources at the within 481 seasonal scale. This tactic appears to be ubiquitous in the roe deer, occurring across its entire 482 483 European distribution, and encompassing a wide gradient of environmental conditions. We suggest that large herbivores may adopt this finer-scale tactic when environmental conditions 484 fluctuate spatially and temporally, independently of seasonal variations. In the present context 485 of climate change, predictions of more frequent and intense climatic events (IPCC 2014) may 486 mean that an increasing number of large herbivore populations adopt the multi-range tactic, 487

488 combining more frequent spasmodic movements interspersed with short periods of489 sedentarism, to track available resources.

490

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503

504 **Ethical approval:** All applicable institutional and/or national guidelines for the care and 505 use of animals were followed.

506

507 Data accessibility

508 Data used for this study are accessible on EURODEER website (http://eurodeer.org) on 509 demand.

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 120.
- 663

Tables

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 -	20	35	36	27
ŭ	Germany	28	55	30	21
	Šumava National Park – Czech	0	2	1	1
e	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
1	Matese - Italy	0	3	3	0
m	Monte Bondone - Italy	4	4	4	4
n	Rendena Giudicarie – Italy	6	3	6	3
0	Østafjells west - Norway	3	6	5	4
	Total	93	193	190	96

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

644 Figure legends

Fig. 1 Location of study sites where roe deer were monitored using GPS collars. Letters
correspond to the 'study site id' in Table 1, with the corresponding number of individuals in
italics

Fig. 2 Example of the segmentation approach using Lavielle's Method (1999) on the movement trajectory of a migratory individual (individual 1367 monitored in 2008) during the winter and the summer seasons (top) and the corresponding sub-seasonal functional home ranges (sfHR) (bottom). The X-axis indicates the chronological order of successive locations and the Y-axis shows the Net Squared Displacement (NSD). Vertical lines correspond to cutoff points defining the sequential sfHR

Fig. 3 Example of the segmentation approach using Lavielle's Method (1999) on the movement trajectory of a resident individual (individual 2296 monitored in 2013), during the winter and the summer seasons (top) and the corresponding sub-seasonal functional home ranges (sfHR) (bottom). The X-axis indicates the chronological order of successive locations and the Y-axis shows the Net Squared Displacement (NSD). Vertical lines correspond to cutoff points defining the sequential sfHR

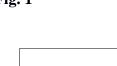
Fig. 4 Graphical representation of the best models describing variation in the number of sub-660 661 seasonal home ranges (sfHR) during the winter season, as a function of the two-way interaction between resource predictability and movement tactic (a) and the log-transformed 662 sfHR size (b); and during the summer season as a function of sex and the log-transformed 663 sfHR size (c). In order to better visualize the raw data, we plotted the mean number of sfHR 664 for each study site. Dotted lines represent the 95% confidence intervals resulting from the 665 666 model predictions, error bars represent the standard deviation in number of sfHR, and the letters correspond to the 'study site id' in Table 1. We fixed movement tactic as resident, 667

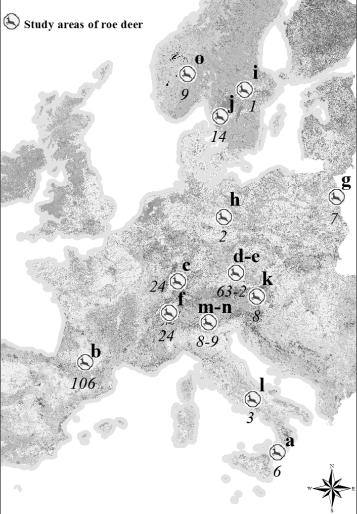
resource predictability and the log-transformed sfHR size at their mean when not representedin the figures

Fig. 5 Graphical representation of the best model describing variation in the log-transformed 670 671 distance among sub-seasonal home ranges (sfHR) during the winter, as a function of the twoway interaction between resource heterogeneity and movement tactic and between resource 672 heterogeneity and sex, represented as a three-way interaction (a: females, b: males), and the 673 674 log-transformed median sfHR size (c). In order to better visualize the raw data, we plotted the 675 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 676 deviation of log-transformed inter-sfHR distance, and the letters correspond to the 'study site 677 id' in Table 1. We fixed resource heterogeneity and the log-transformed sfHR size at their 678 mean, movement tactic as resident and sex as female when not represented in the figures. 679

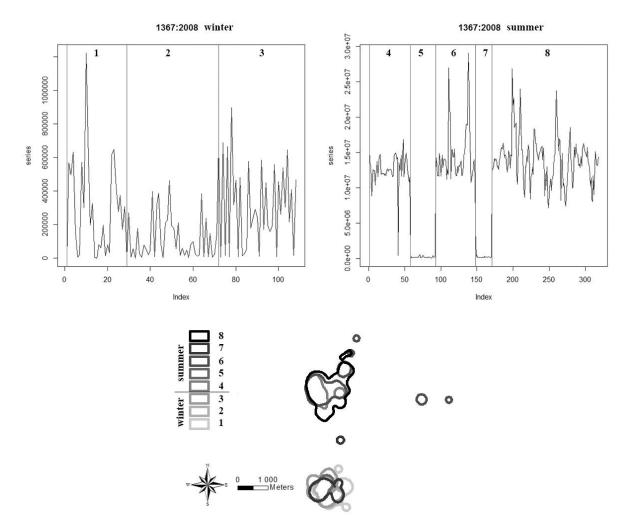
680 Fig. 6 Graphical representation of the best model describing variation in the log-transformed distance among sub-seasonal home ranges (sfHR) during the summer season, as a function of 681 resource predictability (a), sex (b) and the log-transformed median sfHR size (c). In order to 682 better visualize the raw data, we plotted the mean log-transformed sfHR distance for each 683 study site. Dotted lines represent the 95% confidence intervals resulting from the model 684 predictions, error bars represent the standard deviation of log-transformed inter-sfHR 685 distance, and the letters correspond to the 'study site id' in Table 1. We fixed resource 686 predictability and the log-transformed sfHR size at their mean and sex as female when not 687 represented in the figures 688

689

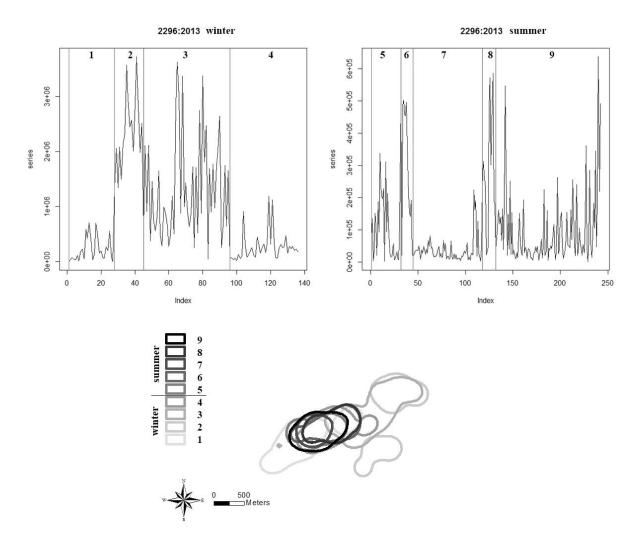




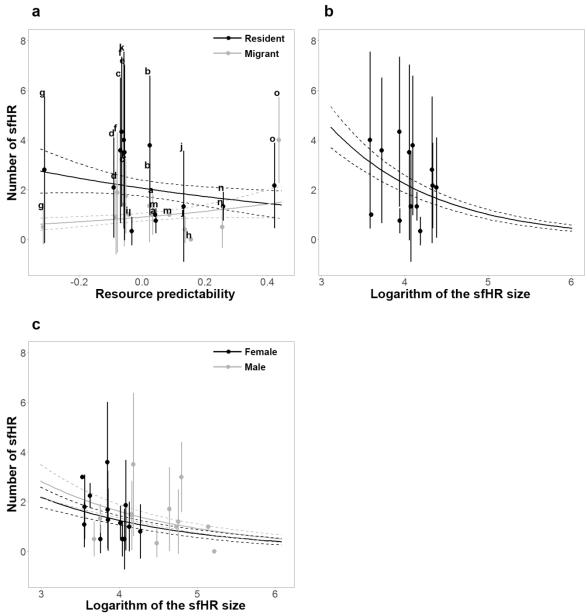












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