A Systematic Review of Within-Population Variation in the Size of Home Range Across Ungulates: What Do We Know After 50 Years of Telemetry Studies?

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Studying the factors determining the sizes of home ranges, based on body mass, feeding style, and sociality level, is a long-standing goal at the intersection of ecology and evolution. Yet, how species-specific life history traits interact with different components of the landscape to shape differences in individual home ranges at within-population level has received much less attention. Here, we review the empirical literature on ungulates to map our knowledge of the relative effects of the key environmental drivers (resource availability, landscape heterogeneity, lethal and non-lethal risks) on the sizes of individual home ranges within a population and assess whether species’ characteristics (body mass, diet, and social structure), account for observed variation in the responses of the sizes of individual home ranges to local environmental drivers. Estimating the sizes of home ranges and measuring environmental variables raise a number of methodological issues, which complicate the comparison of empirical studies. Still, from an ecological point of view, we showed that (1) a majority of papers (75%) supported the habitat productivity hypothesis, (2) the support for the influence of landscape heterogeneity was less pervasive across studies, (3) the response of cattle-type to variation in food availability was stronger than the response of moose-type, and (4) species-specific body mass or sociality level had no detectable effect on the level of support to the biological hypotheses. To our surprise, our systematic review revealed a dearth of studies focusing on the ecological drivers of the variation in the sizes of individual home ranges (only about 1% of articles that dealt with home ranges), especially in the later decade where more focus has been devoted to movement. We encourage researchers to continue providing such results with sufficient sample sizes and robust methodologies, as we still need to fully understand the link between environmental drivers and individual space use while accounting for life-history constraints.

Keywords: ungulate, diet, complementary resource, landscape structure, life history trait, landscape heterogeneity, habitat, body mass
INTRODUCTION

How the size of a home range responds to environmental changes has been the focus of many empirical studies across taxa (“home range,” singular or plural, returns 12,389 hits as a Topic on Web of Science Core Collection by 2 Nov 2020) including many studies on ungulates. Finding the drivers of variation in the sizes of home ranges within a given species across different habitats would indeed allow understanding higher-level processes such as population range expansion and restriction in the context of global changes. The size of an individual’s home range (sensu Burt, 1943; i.e., “area traversed by the individual in its normal activities of food gathering, mating, and caring for young”) should adjust to the different components of the environment, as a result of the movements and habitat selection by this individual (Van Moorter et al., 2016). The relationship between the size of a home range and food resources, which corresponds to the habitat productivity hypothesis proposed by Harestad and Bunnel (1979), is expressed as:

$$H = \frac{R}{P}$$  \hspace{1cm} (1)

with $H$ the size of the home range, $R$ the overall energy requirements of an individual of a given mass (in kilocalories (kcal) per day) and $P$ the rate of usable energy (kcal per day per unit area). Nonetheless, this formulation does not account explicitly for resources other than food (e.g., water, shelter), for avoidance of risk of death, or for social interactions. As habitats are heterogeneous (e.g., Wiens et al., 1985), the size of a home range should also depend on the spatial distribution of the amount and quality of resources (Mitchell and Powell, 2004). This was explicitly mentioned by Harestad and Bunnel (1979) but simply as an allometric constraint of large animals forced indirectly to include non-resource habitats within their home range because of the need to forage over large areas to satisfy all energy requirements. With highly dispersed resources, interstitial patches connecting food plots occur in the habitat matrix. This leads to increased exploitation costs for individuals, hence to a need for larger home ranges (Péron, 2019). While a home range with highly dispersed food patches can still be perceived by an individual as being of high quality (Mitchell and Powell, 2008), the expected relationship between classic estimates of the size of a home range (Laver and Kelly, 2008) and resources within the home range would appear negative. Furthermore, not only food resource distribution matters. Animals also face lethal risks (hunting, predation, and vehicle collision) and non-lethal disturbances perceived as risks (human presence in nature for other reasons than hunting; Ciuti et al., 2012; Berger-Tal and Saltz, 2019) that vary in space and time. These lethal and non-lethal risks generate the so-called landscape of fear (Laundré et al., 2001), which can lead to changes in the sizes of home ranges through, for instance, the need to include more refuge areas (Taylor, 1988; Powell et al., 1996). Lastly, when habitats ensuring different functions (Dunning et al., 1992; Camp et al., 2013) are far from each other, home ranges should increase to include all these habitats. Therefore, the habitat productivity hypothesis (Harestad and Bunnel, 1979), which is only based on food resources for a given mass, appears to be too restricted. Within a population, the relationship between the sizes of home ranges and environmental variables should indeed result from the interplay of individual responses to all types of resources and risks and to landscape heterogeneity. Exploring the complexity of these responses has been the goal of modeling and simulation endeavors (Moorcroft et al., 1999; Mitchell and Powell, 2004; Börger et al., 2008; Buchmann et al., 2013), paralleling the upsurge of empirical studies. After five decades of technological improvements (Wilson et al., 2008; Kays et al., 2015), it is timely to evaluate whether empirical knowledge acquired on a panel of diverse species and environments support hypotheses about the ecological drivers of the size of an individual’s home range.

For a given ecological context, the sizes of home ranges should differ among individuals of different species in relation to body mass and life history tactics (Ofstad et al., 2016). Hence, the overall responses of the sizes of home ranges to changes in environmental features across individuals within a population are framed within these species-specific constraints. The main constraints should result from body mass, which determines energy requirements (see, e.g., Robinson et al., 1983) and the potential distance covered during a unit of time (at least within a taxonomic group), and to life style, which determines resource acquisition (Dobson, 2007). As metabolic rates are hypo-allometrically scaled with body mass, large individuals need less food per mass unit than small ones, so that $R$ in Equation 1 scales with an allometric exponent of ca. 0.75 (Brody, 1945) and are also less selective in terms of food quality than small ones (Demment and Van Soest, 1985). In addition, as habitat features are independent of species size, the available energy per area does not change with body mass (Jetz et al., 2004) but the rate of acquisition does, which leads $P$ in Equation 1 to have the dimension of frequency (with an allometric exponent close to $-0.25$, e.g., Robinson et al., 1983). This difference in allometric exponents leads the size of a home range for a given resource requirement to decrease much faster with increasing resource availability in large than in small individuals, and thereby to expect the decrease of the size of an individual’s home range with increasing $P$ to be weaker in small than in large species.

Among lifestyle traits that determine resource acquisition, diet is of prime importance (Searle and Shipley, 2008) and strongly influences the sizes of home ranges across mammals (Harestad and Bunnel, 1979; Tucker et al., 2014). Ungulates are mostly herbivores with a generalist diet, but they differ in their morpho-physiological characteristics, which cascades into how flexible their diet can be. Codron and Clauss (2010) distinguished species with a “moose-like” rumen, which have a suit of morpho-physiological features that restrict them to feed on browse material, and species with a “cattle-like” rumen, which allows them to have a more flexible diet. Individuals of species with moose-type rumen should therefore be more sensitive to variation in resource availability than individuals with cattle-type rumen, which have a wider diet niche (Codron and Clauss, 2010). In addition, as browse material represents more sparsely distributed resources (Jarman, 1974; Gordon, 2003), the sizes of home ranges of individuals with moose-type rumen,
which are browsers, should be most sensitive to changes in landscape heterogeneity.

Studies performed across species have long pointed out that the sizes of home ranges should also vary with home range overlap among individuals (Jetz et al., 2004; Péron, 2019), and then according to the average group size typical of a species (Damuth, 1981; McLaughlin et al., 2000). Hence, the size of an individual’s home range for a given resource availability should be larger in group-living individuals than in solitary individuals. In populations with diverse group sizes (a common feature of group-living ungulates, Prox and Farine, 2020), the variation in the number of individuals sharing home ranges should weaken the influence of resource availability on the sizes of home ranges.

One last aspect to consider for identifying the relative roles of resources, landscape heterogeneity, and lethal and non-lethal risk on the sizes of home ranges is the seasonality of the resources, as ungulate populations face a succession of productive and non-productive seasons. The signature of food resources on the sizes of home ranges should therefore be stronger in the limiting than in the productive season (Volampeno et al., 2011), even though this effect could be dampened in some species by the ability of individuals to acquire fat reserves during the productive season, allowing them to cope with a limited access to resources during the lean season (Mautz, 1978; Stephenson et al., 2020).

Based on a systematic literature review of the variation in the sizes of individual home ranges at the within-population level in ungulates, we aimed at assessing (1) the empirical support for an impact of the different ecological drivers (resource availability, landscape heterogeneity, and lethal or non-lethal risk) on the size of an individual’s home range and (2) whether the level of support found in the different publications depended on the lifestyle traits (body mass, diet, and social structure) of the species studied. Ungulates occupy all biomes, are central in terrestrial trophic networks (Montgomery et al., 2019), display a large range of both body mass and ecological traits (Fritz and Loison, 2003), and have been the focus of studies on the sizes of home ranges at the individual, population, and species levels since the 1970s (e.g., McNab, 1963; Estes, 1974; Jarman, 1974; Dunn and Gipson, 1977; Owen-Smith, 1977). Thus, ungulate home ranges have been studied by direct observation for decades and by the use of telemetry from the early days of its development for wildlife (e.g., Dunn and Gipson, 1977). Collars with VHF and now GPS and biologgers (Kays et al., 2015) have been deployed on an ever-growing number of species, providing a rich literature on the sizes of home ranges (for interspecific overviews, see, e.g., Mysterud et al., 2001; Ofstad et al., 2016). Yet, to map the determinants of the sizes of home ranges among individuals across species, methodological questions arise about how the sizes of home ranges are estimated from location data, and how environmental variables are measured (Nilsen et al., 2005). These are not trivial issues as attested by the substantial literature focusing on the meaning and definition of a home range (Kie et al., 2010; Fieberg and Börger, 2012; Powell and Mitchell, 2012; Péron, 2019), as well as on statistical methods to estimate not only the sizes of home ranges (Börger et al., 2006a; Kie et al., 2010; Noonan et al., 2019; Péron, 2019) but also resource availability (from the field, Flombaum and Sala, 2007; Redjadj et al., 2012; or remotely, Pettorelli et al., 2005, 2006; Garrouste et al., 2016; and more specifically, the edible portion of the biomass, Duparc et al., 2020) and landscape heterogeneity (Sundell-Turner and Rodewald, 2008).

In the midst of these ecological and methodological challenges, we reviewed the empirical support for the impact of three main ecological drivers (resources, landscape heterogeneity, and risks), on the variation in the sizes of individual home ranges, accounting for article-specific confounding methodological variables (sample size, positioning system, number of locations, number of metrics considered, home range estimator). Comparing the results across all articles, we then tested whether the effect of resources on the sizes of home ranges was supported more pervasively during the productive than during the non-productive season. Furthermore, we checked whether lifestyle traits such as body mass, group size and diet explained the level of support to the ecological hypotheses found in published articles (Figure 1). After mapping the state of our knowledge after five decades of upsurge in the number of studies on the sizes of home ranges in ungulates, and in times when new technologies revolutionize the fine-scale monitoring of animal movement and behavior (Wilson et al., 2008, Kays et al., 2015), we identify the remaining challenges and raise guidelines for future studies.

MATERIALS AND METHODS

Literature Survey

We conducted the literature survey using the Web of Science Core Collection, and then applied the PRISMA procedure (Liberati et al., 2009, Figure 2) to select a collection of comparable papers (Figure 1). We used the following keywords (“home range” OR “range use” OR “spatial distribution” OR “space use” OR “spatial variation” OR “area use” OR “habitat use” OR “ranging behaviour”) AND (“ungulate” OR “herbivore” OR “mammal” OR “deer” OR “chamois” OR “ibex” OR “mouflon” OR “bison” OR “capreolus”) and obtained a total of 7,454 articles (Figure 2). We deliberately added a few names of species as keywords to incorporate older studies, which usually used the name of species in titles or keywords instead of broader taxonomic (e.g., ungulates) or ecological (e.g., herbivores) terms. We restricted the results of our search to the topics Ecology, Zoology, Environmental, and Behavioral sciences. The survey included articles published up to October 2019. Our selection followed three steps (Figure 2). We first retained for further consideration only articles for which data on the sizes of home ranges were provided for ungulates and at the population level. We restricted our review to herbivorous ungulates, hence removing articles on the wild boar (Sus scrofa), an omnivorous species, but included articles on elephant (Loxodonta africana), which is an herbivorous subungulate. This led us to keep 216 articles out of the 7,454. Second, we retained only articles providing a continuous relationship between the sizes of individual home ranges and an independent ecological metric measured at the individual home range level. This led us to remove 165 and retain 53 articles. Two of the discarded articles (Börger et al., 2006b on roe deer, Capreolus capreolus, and Brook, 2010 on elk, Cervus
canadensis) provided the sizes of home ranges for individuals assigned a priori to habitat categories defined by their dominant vegetation type, which were synthetic proxies of differences in all ecological drivers (resource, landscape heterogeneity, risk). We could therefore not derive results concerning the relationship between the size of an individual’s home range and an ecological metric of resource, landscape heterogeneity or risk from these articles that would compare with the other retained articles. Third, we considered only articles that provided the sizes of seasonal home ranges to test the hypothesis that the expected effect of resources on the sizes of individual home ranges should get more support during the non-productive than during the productive season. Although we recognize that there are several meanings and definitions of home range in the literature (see introduction), here we have had to accept by default the definition adopted by the authors in each of the selected studies.

**Literature Analysis and Retrieval of Article and Species Metadata**

For each retained article, we started by recording the basic information required to identify each article (i.e., title, first author, year of publication, journal), the species studied and study sites. When an article considered several species or several study sites, we split it accordingly, as our unit of interest for retrieving the environmental-to-home range relationship was the species-specific population. Then, we called a “case-study” each unique species-population, which became our unit of study. We retrieved data related to animal sampling (sample size, age, and sex), duration and period of monitoring (e.g., all year round,
during which season) (Table 1). We also recorded information about the location data: type of collars used (VHF or GPS) and number of fixes used per animal to estimate the size of its home range. Then, we classified each paper according to the method used to estimate the sizes of home ranges [Minimum Convex Polygon (MCP), Kernel Density Estimation (KDE), Local Convex Hull (LoCoH), Harmonic Mean (HM), and Brownian Bridge (BB)], thereafter regrouped into three classes: MCP, KDE vs. other methods for further analysis. We also classified papers according to time frame (i.e., day, week, month, season, or year), and spatial scale in terms of isopleths (e.g., 50 or 90% of utilization distribution). We discarded estimates at a time frame lower than 1 month, because the ecological meaning of very short-term home ranges is debatable (Péron, 2019), and at annual time frame because we were interested to test the impact of seasonal variation in resources on the support of ecological drivers. It is worth noting that the few articles estimating the sizes of home ranges at the day or week levels also provided estimates at a longer time frame, so we did not discard any case study for this reason. We did not retain sizes of home ranges estimated with <70% of locations.

As a second step, we listed all the ecological variables studied in a case study. For each of these variables, we retrieved the unit of measurement, the spatial scale at which it was estimated (e.g., at the home range scale or in a buffer around the home range), and whether it was log-transformed before being included in the analyses. Indeed, the relationship between the sizes of home ranges and productivity becomes linear only when both of them are log-transformed (Equation 1). We faced a huge diversity of metrics (Supplementary Table 1), which we classified as measures of resource, landscape heterogeneity, or risk (Figure 1; Supplementary Table 1). We then counted the number of metrics for a given category of variables (resource, landscape heterogeneity, risk) and a given season (productive and non-productive season) studied per case study. We did not consider the following metrics: index of snow severity (Ramanzin et al., 2007), snow cover (Grignolio et al., 2003), snow depth (van Beest et al., 2011; Scillitani et al., 2012), elevation (Hansen...
### TABLE 1: Information on the 33 selected papers (productive and non-productive seasons).

| References                  | Study duration (year) | Comment | n  | Species          | Sex      | Resource | Risk | Landscape heterogeneity |
|-----------------------------|-----------------------|---------|----|------------------|----------|----------|------|------------------------|
| Anderson et al. (2005)      | 4                     | Population | 10 | Elk              | Female   | x        |      |                        |
| Anderson et al. (2005)      | 4                     | Population | 32 | Elk              | Female   | x        |      |                        |
| Anderson et al. (2005)      | 4                     | Population | 57 | Elk              | Female   | x        |      |                        |
| Bartlam-Brooks et al. (2013)| 1                     | 14      | Zebra | Female | x | x |      |                        |
| Bender et al. (2007)        | 2                     | 40      | Mule deer | Female | x |      |      |                        |
| Bevanda et al. (2015)       | 8                     | Species | 32 | Red deer         | Combined | x |      |                        |
| Bevanda et al. (2015)       | 6                     | Species | 40 | Roe deer         | Combined | x |      |                        |
| Bjørneraas et al. (2012)    | 2                     | Sex     | 108 | Moose            | Female   | x |      |                        |
| Bjørneraas et al. (2012)    | 2                     | Sex     | 108 | Moose            | Male     | x |      |                        |
| Braschares and Arcese (2002)| 0.5                   | 161     | Oribi | Female | x |      |      |                        |
| Hansen et al. (2009a)       | 2                     | 29      | Svalbard reindeer | Female | x |      |      |                        |
| Hansen et al. (2009a)       | 2                     | 26      | Svalbard reindeer | Female | x |      |      |                        |
| Daleszczyk et al. (2007)    | 11                    | Population | 9 | European bison | Combined | x |      |                        |
| Daleszczyk et al. (2007)    | 11                    | Population | 25 | European bison | Combined | x |      |                        |
| de Beer and van Aarde (2008)| 2                     | Population | 6 | African bush elephant | Female | x | x |                        |
| de Beer and van Aarde (2008)| 2                     | Population | 4 | African bush elephant | Female | x | x |                        |
| de Beer and van Aarde (2008)| 2                     | Population | 4 | African bush elephant | Female | x |      |                        |
| Dussault et al. (2005)      | 2                     | 8       | Moose | Combined | x |      |      |                        |
| Granger et al. (2005)       | 21                    | Sex     | 7  | African bush elephant | Male | x | x |                        |
| Granger et al. (2005)       | 21                    | Sex     | 8  | African bush elephant | Female | x | x |                        |
| Grignolio et al. (2003)     | 2                     | 14      | Alpine ibex | Male | x |      |      |                        |
| Kilpatrick et al. (2011)    | 2                     | 56      | White-tailed deer | Female | x | x | x |                        |
| Lamberti et al. (2006)      | 1                     | Sex     | 4  | Roe deer         | Female   | x |      |                        |
| Lamberti et al. (2006)      | 1                     | Sex     | 9  | Roe deer         | Combined | x |      |                        |
| Lamberti et al. (2006)      | 1                     | Sex     | 5  | Roe deer         | Male     | x |      |                        |
| Laurian et al. (2008)       | 3                     | 47      | Moose | Combined | x |      |      |                        |
| Leach and Edge (1994)       | 2                     | 13      | White-tailed deer | Female | x |      |      |                        |
| Massé and Côté (2012)       | 5                     | Season | 19 | White-tailed deer | Female | x | x |                        |
| Massé and Côté (2012)       | 5                     | Season | 13 | White-tailed deer | Female | x | x |                        |
| Moe and Wegge (1994)        | 2                     | 10      | Axis deer | Female | x |      |      |                        |
| Morellet et al. (2013)      | 1.5-8                 | 190     | Roe deer | Female | x |      |      |                        |
| Naidoo et al. (2012)        | 3                     | 31      | African buffalo | Combined | x | x |      |                        |
| Nicholson et al. (1997)     | 3                     | 43      | Mule deer | Combined | x | x |      |                        |
| Quinn et al. (2013)         | 2                     | Season/sex | 32 | White-tailed deer | Female | x |      |                        |
| Quinn et al. (2013)         | 2                     | Season/sex | 15 | White-tailed deer | Male | x |      |                        |
| Quinn et al. (2013)         | 2                     | Season/sex | 33 | White-tailed deer | Female | x |      |                        |
| Quinn et al. (2013)         | 2                     | Season/sex | 17 | White-tailed deer | Male | x |      |                        |
| Rivrud et al. (2010)        | 3                     | 47      | Red deer | Female | x |      |      |                        |
| Said and Servanty (2005)    | 2                     | 24      | Roe deer | Female | x |      |      |                        |
| Said et al. (2005)          | 2                     | 37      | Roe deer | Female | x |      |      |                        |
| Scillitani et al. (2012)    | 6                     | 28      | Alpine ibex | Male | x |      |      |                        |
| Tufto et al. (1996)         | 5                     | 35      | Roe deer | Female | x |      |      |                        |
| van Beest et al. (2011)     | 2                     | 24      | Moose | Female | x |      |      |                        |
| Vercauteren and Hygstrom (1998)| 3             | 30      | White-tailed deer | Female | x |      |      |                        |
| Viana et al. (2018)         | 3                     | 18      | Iberian ibex | Combined | x |      |      |                        |
| Walter et al. (2009)        | 7                     | 257     | White-tailed deer | Female | x | x | x |                        |
| Widmer et al. (2004)        | 2                     | 6       | Roe deer | Female | x |      |      |                        |

The column “comment” indicates the study of several populations. The last three columns indicate the category of variables tested in each paper (resource, landscape heterogeneity, and risk; see Figure 1).
et al., 2009a; Bevanda et al., 2015), and slope (Anderson et al., 2005), since these factors were only relevant to some specific environments (e.g., mountainous and arctic regions). Based on the species’ ecology and information given in each article, we associated each metric to the expected slope direction of its relationship with the sizes of home ranges: increase, i.e., positive effect size, or decrease, i.e., negative effect size. For example, to be in line with the resource hypothesis, we expected a negative relationship between the sizes of individual home ranges and resource metrics reflecting good quality forage (e.g., grass nitrogen content, Brashares and Arcese, 2002), but a negative relationship when resource metrics were related to poor-quality forage (e.g., fiber content, Brashares and Arcese, 2002).

As a third step, we retrieved species-specific traits from the literature, such as adult body mass for both sexes, type of physio-digestive system (moose-type vs. cattle-type), sociality level, and phylogenetic information (see Table 2). We did not use phylogenetic information in our models due to the low number of species (15 species) finally retrieved in our sample (see “Results” section). For ranking species on the generalist-specialist gradient, we used the type of physio-digestive system (moose-type vs. cattle-type), as proposed by Codron and Clauss (2010). For the social structure, we classified species as living in small groups (solitary to small groups up to five individuals) and in large groups (six or more), following the classification proposed by Prox and Farine (2020).

### Extraction of Results

In the retained articles, independent variables were considered to influence the sizes of home ranges either based on an information theory approach [usually comparing models with an information criterion, such as Akaike Information Criterion (AIC)] or by inferential procedures based on statistical tests and p-values. Retrieving standardized effect sizes required for a meta-analysis was not possible because independent variables were mostly included in multivariable models, were not systematically scaled or log-transformed, and statistics needed to calculate effect sizes were not all (or not at all) provided in the article (especially for variables found to have no statistically significant effect or that were not included in the best models). Consequently, we summarized the results in a categorical way instead of reporting the estimated values of effect sizes (i.e., slopes and standard errors). For each metric (Figure 1), we recorded, whether the analyses reported in the article supported our biological hypotheses regarding resource, landscape heterogeneity, or risk, i.e., whether the relationship between the sizes of home ranges and the ecological variable was in the same direction as expected, in the opposite direction, or not retained. For independent variables without detectable effect or not included in the selected model, we sought whether the slope was nevertheless in the expected direction, or opposite to the expected direction, when given. Therefore, we ended up with variables assigned to five different result categories: “as expected and detected or retained in the best models”, “as expected and not detected or not retained in the best model”, “opposite and detected or retained in the best model”, “opposite and not detected or not retained in the best model”, or “direction unknown and not detected or not retained in the best model”. The distribution of the five different categories of results is provided in Supplementary Figure 1. For the rest of our analyses, we classified the result of each metric as being supporting, or not supporting, the biological hypothesis it was assigned to Figure 1. Importantly, we considered that a biological hypothesis was supported in a case study if at least one of its metrics was supporting the given biological hypothesis.

### Table 2: Characteristics of the species included in the 33 selected papers (productive and non-productive seasons).

| Species                        | Latin name         | ♂ BM (kg) | ♀ BM (kg) | Group size | Habitat                      | Diet           |
|--------------------------------|--------------------|----------|----------|------------|------------------------------|----------------|
| Moose                          | Alces alces        | 440      | 320      | 4          | Boreal and mixed forest      | Moose-type     |
| Axis deer                      | Axis axis          | 80       | 55       | 25         | Grassland forest             | Cattle-type    |
| European bison                 | Bison bonasus      | 710      | 420      | 20         | Mixed forests                | Cattle-type    |
| Ibex                           | Capra ibex         | 95       | 45       | 11         | Alpine grassland             | Cattle-type    |
| Iberian ibex                   | Capra pyrenaica    | 70       | 35       | 7          | Shrubland areas              | Cattle-type    |
| Roe deer                       | Capreolus capreolus| 26       | 26       | 2          | Mixed                       | Moose-type     |
| Elk                            | Cervus canadensis  | 312      | 238      | 4          | Grassland                   | Cattle-type    |
| Red deer                       | Cervus elaphus     | 163      | 107      | 6          | Mixed                       | Cattle-type    |
| Zebra*                         | Equus burchelli    | 280      | 190      | 10         | Grassland-savannah           | Cattle-type    |
| African bush elephant*         | Loxodonta africana | 8000     | 3700     | 12         | Dry wood/shrublands          | Moose-type     |
| Mule deer                      | Odocoileus hemionus| 80       | 50       | 3          | Mixed                       | Moose-type     |
| White-tailed deer              | Odocoileus virginianus| 72     | 48       | 3          | Mixed                       | Moose-type     |
| Oribi                          | Ourebia ourebi     | 14       | 13       | 2          | Mixed                       | Cattle-type    |
| Svalbard reindeer              | Rangifer tarandus  | 136      | 88       | 6          | Open taiga                  | Cattle-type    |
| African buffalo                | Syncerus caffer    | 700      | 500      | 30         | Savannah                    | Cattle-type    |

Species are ranked by Latin name. See Supplementary Table 2 for references. *These two species are not ruminants, and therefore should not be classified following the “moose-type” and “cattle-type” typology of Codron and Clauss (2010). Nevertheless, the respective diets of these species, mixed-feeder for elephant and grazer for zebra, correspond to the use of resources of moose-type and cattle-type species, respectively, with strong physiological differences (McNaughton and Georghiads, 1988).
independently of the number of metrics tested. We are aware that a statistical hypothesis can only be rejected or not rejected, but for greater clarity throughout the manuscript, we use the terms "supporting" the biological and ecological hypothesis tested for "rejecting" the null statistical hypothesis, and "not supporting" the biological and ecological hypothesis tested for "not rejecting" the null statistical hypothesis.

Descriptive and Statistical Analysis
We performed an in-depth analysis of the probability for a study to support each of the biological hypotheses. Using generalized linear models (GLMs), we first tested whether methodological covariates influenced the probability of support for a given biological hypothesis, considered as a binomial response variable (Figure 1). The methodological variables tested were sample size (for a given expected effect size, large sample size should positively influence the likelihood to detect a relationship), the number of variables studied (the larger the number of variables considered, the more likely one would be retained), and all technical aspects expected to increase the accuracy of estimates of the sizes of individual home ranges (number of relocations; positioning system, VHF vs. GPS; statistical method used to estimate the sizes of home ranges). We then assessed if the level of support for a biological hypothesis differed for resource, landscape heterogeneity, or risk and whether it was more important for resources during the non-productive season than during the productive season. Note that due to the few case studies focusing on risk (and none on risk only), we did not include this category in models, hence we were not able to assess the level of empirical support for our hypothesis that individual exposure to risk should affect the sizes of individual home ranges. We included a two-way interaction between the ecological driver (resource and landscape heterogeneity) and the number of metrics tested, as this number was generally higher for landscape heterogeneity. We compared models accounting for methodological variables using the Akaike Information Criterion and selected the model with the lowest AIC value.

Then, we assessed whether the lifestyle variables (Figure 1) could influence the level of support to each ecological driver: body mass (expecting an increased support with increasing body mass), diet (expecting a stronger support for species with moose-type digestive features), and social structure (expecting a reduced support in large group-living species).

We fitted GLMs with a binomial error structure and a logit link function, using the "glm" function implemented in the package "stats" of R 3.3.3 (R Development Core Team, 2017). We ranked candidate models using the Akaike Information Criterion for small sample sizes (AICc) as implemented in the R package "AICcmodavg" (Mazerolle, 2020) and calculated ΔAICc and AICc weights. Models with ΔAICc ≤ 2 were considered equivalent (Burnham and Anderson, 2002), and when models had a ΔAICc ≤ 2, we kept the most parsimonious one. Results are depicted using the "sjPlot" (Lüdecke, 2019) and "visreg" (Breheny and Burchett, 2017) R-packages.

RESULTS

Studies and Species
We retained 33 articles, published in 26 different scientific journals that met our two first criteria of selection (Figure 2), covering both the productive (29 articles) and non-productive (19 articles) seasons. Most studies were published during the early 2000–2010, with a noticeable decrease post-2013 (Figure 3A). Fifteen species were studied during at least one of the productive and non-productive seasons (Figure 3B). These species belong mostly to the order Artiodactyla, and are essentially members of the families Bovidae (five species) and Cervidae (eight species)—the remaining two species belong to the Equidae (Equus burchelli) and Elephantidae (Loxodonta africana) (Figure 3B). Among these species, some have been studied more than others, as 19 of the 33 articles (57%) focused on only three species: roe deer Capreolus capreolus (eight articles), white-tailed deer Odocoileus virginianus (seven articles), and moose Alces alces (four articles). Individuals from different species were contrasted in terms of body mass, group size, habitat, and percentage of grass in the diet (mean = 45.9 ± 29.5): six species were classified as having individuals forming low group sizes and nine as having highly social individuals (Table 2).

Almost all studies have been conducted in North America (n = 10) and Europe (n = 17), with only five studies located in Southern Africa (Figure 3C). We retrieved only one study conducted in Asia [axis deer Axis axis in Nepal, Moe and Wegge (1994)]. Among the 33 retained articles, three included several populations of the same species (Anderson et al., 2005; Daleszczyk et al., 2007; de Beer and van Aarde, 2008) and one article studied two species (Bevanda et al., 2015). Overall, this led to 39 independent case studies (Table 1).

Telemetry and Estimators of the Sizes of Home Ranges
Animal locations were obtained either by GPS or VHF positioning system. Before 2000, all studies used VHF collars (Figure 4A). GPS collars came into use in 2000 and dominated until after 2010 (Figure 4A). Before 2000 (five case studies), MCP dominated, and two methods (KDE and Harmonic Mean method as the only “other methods”) were used equivalently (Figure 4B). During the 2001–2010 (23 case studies), KDE took over MCP, with 63% of articles using KDE. Since 2011 (11 case studies), methods used have diversified, leading to a decrease in the occurrence of MCP and KDE. The increased use of other methods in the later years corresponded to the development of alternative home range estimators such as LoCoH or brownian bridge methods. Twenty-five case studies (65%) were based on >20 marked animals, while only six case studies (15%) had sample sizes above 50 (Figure 4C). The number of relocations used to estimate the sizes of home ranges also revealed a strong variability among the 39 case studies (Figure 4D). On average, about 852 locations per month were used to compute the sizes of home ranges, but with a wide range, with minimum and maximum numbers of locations being 6 and 5,040, respectively.
Descriptive Summary of Studied Variables
Most case studies focused on variables related to “resource” (n = 34), followed by “landscape heterogeneity” (n = 12) and lethal or non-lethal “risk” (n = 6; Figure 5A). No case study focused solely on lethal or non-lethal risk, and this factor was always tested in conjunction with one of the other two categories of variables or with both of them. All variables considered as metrics of risk by the authors were distance or density of linear features (e.g., road) (Supplementary Table 1). The majority of case studies investigated the effect of several independent variables on the sizes of home ranges with an average 2.6 relationships tested per case study for resource (range: 1–12), 7.8 for landscape heterogeneity (range: 1–12), and 1.3 for risk (range: 1–3) (Figure 5B). As too few studies assessed the influence of lethal or non-lethal risk, this ecological driver was excluded from the analyses. We retrieved 123 slopes characterizing the link between the sizes of home ranges and resource, 152 for the category of landscape heterogeneity variables, and eight for risk metrics. Only 15 papers out of the 33 log-transformed the sizes of home ranges.
The results at the metric level supported one of the biological hypotheses and were detected or retained for 37% of the metrics across all studies, whereas they were opposite to the initial expectation and detected or retained for 13% of the metrics (Supplementary Figure 1). Not detected or non-retained metrics were reported for 50% of the metrics (16% were consistent and 16% opposite to the direction expected under the biological hypotheses and finally 18% were unknown, Supplementary Figure 1).

Methodological Covariables and Seasonal Variation Influencing Papers’ Results

Eighty percent of the 33 case studies focusing on resource metrics (28 in the productive season and 20 in the non-productive seasons) concluded that the habitat productivity hypothesis was supported (i.e., at least one variable tested supported the expected impact of resources on the sizes of home ranges). This percentage was 73% for the 12 case studies studying landscape heterogeneity (9 in the productive season, and 10 in the non-productive seasons; Figure 5C).

The probability for a case study to support the biological hypotheses increased with sample size ($\beta = 0.23 \pm 0.30$, log scale; Figure 6A) and was influenced by the two-way interaction between the number of metrics and the biological hypothesis considered ($\beta = 1.21 \pm 0.52$ for resource, $\beta = -0.07 \pm 2.02$ for landscape heterogeneity; Figure 6B) and by the positioning system ($\beta_{VHF - GPS} = 1.51 \pm 1.27$; Figure 6C and see Supplementary Table 3 for details). Neither the number of locations per individual nor the estimation method of the sizes of home ranges (KDE, MCP, and others) was retained in the best models (all $\Delta$AICc > 2; see Supplementary Table 3). When looking at the prediction from the best model, the percentage of studies supporting the expected relationship between the sizes of home ranges and resources were close to 100%, when animals...
FIGURE 5 | Venn diagrams for (A) the number of case studies testing a relationship between the sizes of home ranges and resource, landscape heterogeneity or risk indices, and combinations of these. (B) Number of metrics tested per case study, per period, and for each ecological driver. (C) Proportion of case studies that contained at least one metric supporting the biological hypothesis, for each ecological driver and season.

were monitored by VHF or by GPS (mostly from papers after 2010, Figure 3), as long as enough resource metrics (roughly >4) were measured (Figure 6B). Landscape heterogeneity influenced the sizes of home ranges in about 96% of studies based on individuals monitored by VHF but in only about 75% for individuals monitored by GPS, whatever the number of studied metrics. Neither the season nor the interaction between season and category of variables were retained in the best model for resources (Supplementary Table 3).

Influence of Species-Specific Life History Traits on Findings

We added body mass (log-transformed), group size, and diet to the baseline model selected above to test for the lifestyle-related hypotheses (Figure 1). Body mass, group size, or diet did not affect the level of support for the biological hypothesis in models where both resource and landscape heterogeneity were considered together (Supplementary Table 4). Only the variables of the baseline model (i.e., sample size, positioning system, and the two-way interaction between the number of metrics and the ecological driver) were retained.

We then focused on the metrics from the resource category only to test more specifically the resource hypothesis and whether the support for the resource hypothesis was larger in species with a moose-type than a cattle-type rumen (model selection table in Supplementary Table 5). We actually found the opposite, as the support to the resource hypothesis was larger for species with a cattle-type rumen ($\beta = 0.81 \pm 0.62$; Figure 6D) than for moose-type ($\beta = -1.48 \pm 0.43$).

DISCUSSION

The size of an individual's home range should vary according to a triptych of factors composed by resources, landscape heterogeneity, and risk (Desy et al., 1990; Haskell et al., 2002; Kittle et al., 2008; Bonnot et al., 2013; Creel et al., 2014). Assessing the level of support to each of these drivers of variation in the sizes of individual home ranges from a literature review has however proven more challenging than we had anticipated, for many reasons. First, a dearth of studies provided exploitable results on the variation of the sizes of home ranges with ecological drivers (<1%). Second, methodological issues included the rampant debate on how to estimate both the sizes of home ranges and independent metrics, issues which are far from trivial. Third, the limited number of species studied and the limited geographic distribution of study sites, tied with the methodological concerns, are hurdles in our ability to test whether species-specific lifestyles explain the different sensitivity of individuals from each species to the main ecological drivers.

More Support to the Resource Than to the Landscape Heterogeneity Hypothesis

Resource metrics, especially food resources, have been studied more than metrics of landscape heterogeneity (Figure 5A). The majority of articles found that the sizes of individual home ranges were decreasing with increasing amounts of resources (Figure 5C), supporting the Habitat Productivity Hypothesis. On the other hand, support for the effect of landscape heterogeneity on the sizes of home ranges was more limited (Figure 5C). Landscape heterogeneity includes several components (local diversity, landscape complementarity, or landscape-level fragmentation; Dunning et al., 1992), each of which should be measured explicitly (Turner, 2005). The lower proportion of papers supporting that landscape heterogeneity affects the sizes of home ranges may therefore result from the difficulty to identify and measure the proper metric and scale at which it should be measured (Bevanda et al., 2015).
Study Design and Statistical Considerations

Sample Size and the Problem of Statistical Power

Articles were more likely to support the biological hypotheses when the studies included a large number of animals (Figure 6A), due to a higher statistical power (Steidl et al., 1997). Interestingly, the effect size linking environmental variables and the sizes of home ranges varied with life history traits (Figure 6D). Authors should therefore anticipate the statistical power of their study accounting for the biology of individuals of the studied species and their life history, to enhance the probability of finding an effect. Biological conclusions based on small samples should not be over-interpreted to avoid erroneous conclusions (Steidl et al., 1997). Fortunately, as the cost of telemetry collars are decreasing, we can expect the number of animals per study to increase for a given budget.

Technological Improvements and Estimates of the Sizes of Home Ranges

The technology to monitor individuals has greatly improved during the last 30 years (Kays et al., 2015) and, accordingly, the
publications after 2010 rely mostly on GPS-collared individuals. GPS collars are more expensive than VHF collars but provide a larger number of predetermined fixes, which should ensure a better delimitation and estimation of the size of a home range (Laver and Kelly, 2008; Pellerin et al., 2008; Kie et al., 2010). Yet, studies relying on GPS collars supported biological hypotheses to a lower extent than studies where animals were VHF collared. A tendency to publish non-significant results to a greater extent in more recent years might account for this surprising result. In parallel to the change in technology, methods to estimate the sizes of home ranges have shifted from MCP to a large range of methods, which undermines the comparison and repeatability of studies (Laver and Kelly, 2008). While we did not detect any effect of the estimation method, the low number of case studies prevented us from testing for all interactions, leading us to interpret this finding with caution. Recent articles studying habitat selection or animal movements tend to skip providing the sizes of home ranges, focusing on smaller spatial scale movement behavior. This probably explains why, despite recent evaluation of the robustness of estimators of the sizes of home ranges (e.g., Noonan et al., 2019), we have retrieved fewer articles and data over the last few decades than we had hoped.

Environmental Variables: Their Number, Measurement, and Meaning

The large diversity of metrics used to measure resource availability and landscape heterogeneity, as well as the restricted number of variables assessing risk, were problematic (Supplementary Table 1). The few risk-related metrics interpreted by the authors to be risky were mostly landscape features (see Table 1) and almost no studies fitted our selection criteria for evaluating individual exposure to natural predation or hunting risks (Figure 1). The complexity to evaluate risk metrics at the individual level renders empirical studies of the possible effect of both lethal and non-lethal risk on the sizes of individual home ranges particularly challenging.

The proportion of relationships supporting biological hypotheses was quite low: 52% with resource metrics and only 21% with landscape heterogeneity metrics (Supplementary Figure 2). Quantifying resources that are edible for ungulates is tricky as it requires information on individual energetic requirements and nutritional state, diet, and diet selection (Duparc et al., 2020). Metrics used to measure these factors, derived from remote sensing, field measurements, and proportion of rich or poor habitat available, are often quite crude (Pettorelli, 2013), far from the species-specific foodscape (sensu Searle et al., 2007; Duparc et al., 2020). Likewise, the fragmentation perceived by an individual (Sundell-Turner and Rodewald, 2008) is difficult to measure from geographic databases where habitats are human-derived categories of land cover and would not necessarily reflect how animals perceive their surrounding landscape (Li and Reynolds, 1994; Baguette and Van Dyck, 2007). These difficulties to measure meaningful metrics may explain also why the probability for a case study to conclude opposite to the expectation (i.e., at least one metric tested contradicting the initial hypothesis) was quite high. The papers with such surprising results were mainly those with a large number of metrics tested. With the continuing increases in map resolution and the growing easiness to analyze patterns in a landscape through dedicated software and packages (e.g., Fragstats, Fourier transforms, Rocchini et al., 2013), researchers are tempted to include a plethora of variables, which is not advisable (Streiner and Norman, 2011). This problem probably explains the low proportion of variables found to be statistically significant, the high probability of obtaining at least one variable with an effect opposite to what is expected, and the ad hoc explanation of significant relationships. A better understanding of the energetic requirements, diet needs, and the multiple components of landscape heterogeneity as perceived by an individual is badly required and should be the focus of renewed empirical effort in the future.

Temporal and Spatial Scales

Seasonality

Contrary to our expectations, the influence of resources on variation in the sizes of home range was not supported to a greater extent during the non-productive than during the productive season. This result was surprising, given that most studies took place in areas with marked seasonality (Figure 3C). Environmental constraints might account for this discrepancy, as movements can be restricted during winter, especially at high latitude and high elevation, preventing the sizes of individual home ranges from responding to reduced food resources during the lean season (Mautz, 1978), so that an extension of the size of the home range may not be required. Hence, while seasonal expansion-contraction of the sizes of home ranges occurs in ungulates (Börger et al., 2006b; Morellet et al., 2013), among-individual variation in home range sizes does not appear to be triggered by food resources to a greater extent in the lean season.

Temporal Definition of Seasonal Period

Another reason why seasonal differences did not affect home range sizes may be that estimates of the sizes of home ranges were calculated over varying time windows. For the sake of comparison, we simply attributed a season to an estimate of the size of a home range, but this classification may itself cover various periods of time, from the peak period of productivity (i.e., within a few weeks), to the productive season on its whole (i.e., over several months). The estimates of both the sizes of home ranges (Börger et al., 2008) and the metrics of resources may vary with the time window. For instance, van Beest et al. (2011) reported that the lack of consistency in the effects of individual attributes and environmental conditions (resource and climate) on the sizes of home ranges could be due to the inadequate choice of the temporal scale (Table 3). A relevant choice of spatial and temporal scales is essential and should be suited to the biology of the species, in terms of how individuals perceive the environment (Cushman and McGarigal, 2004) and stationarity (Laver and Kelly, 2008; Péron, 2019).
TABLE 3 | Biological interpretations provided by authors to explain results that were retained as opposite to initial biological hypothesis, per category of variable (resource availability and landscape heterogeneity).

| Variables                        | Expected | Interpretations given by authors                                | Species              |
|----------------------------------|----------|-----------------------------------------------------------------|----------------------|
| Resources                        |          |                                                                  |                      |
| Proportion of grassland          | Negative | Complementarity (animals searched cover or refuge)               | Mule deer            |
| Proportion of crop               | Negative |                                                                  | White-tailed deer    |
| Proportion of open area          | Negative |                                                                  | Axis deer            |
| Browse density                   | Positive | Scale, disturbance period                                       | Moose                |
| NDVI band                        | Positive | Quality (difference between poor and rich habitat, generalist search quality in rich areas and quantity in poor areas) | Svalbard reindeer    |
| Grass fiber content              | Negative | Small ruminant look for low rate of fiber to digest faster       | Orbi                 |
| Proportion of barren land        | Positive | Low productivity                                                | Moose                |
| Mean foraging biomass            | Negative | None                                                            | Elk                  |
| Proportion of deciduous forest   | Negative | None                                                            | Red deer, Roe deer   |
| Proportion of acacia woodland    | Positive | None                                                            | Zebra                |
| Proportion of lacustrine woodland| Positive | None                                                            | Zebra                |
| NDVI                            | Negative | None                                                            | Svalbard reindeer    |
| Distance to forest               | Negative | None                                                            | White-tailed deer    |
| Distance to water                | Positive | None                                                            | Mule deer            |
| Proportion of low-productivity coniferous forest | Negative | None                                                            | Moose                |
| Vegetation index                | Negative | None                                                            | African buffalo      |
| Landscape heterogeneity          |          |                                                                  |                      |
| Patch number                     | Positive | Many patches in a poor quality environment can cause an increase of home range | Zebra                |
| Interspersion-juxtaposition index| Positive | None                                                            | Zebra                |
| Largest patch index              | Positive | None                                                            | African bush elephant|
| Mean edge contrast               | Negative | None                                                            | African bush elephant|
| Patch density of developed types | Positive | None                                                            | White-tailed deer    |
| Shannon diversity index          | Negative | None                                                            | White-tailed deer    |

Accounting for Life History Traits

Body Mass

Species’ body mass did not influence how the sizes of home ranges responded to environmental factors, despite the range of body mass varying from 13 to 8,000 kg, thus covering the range of body mass of ungulates (Fritz and Loison, 2003). We expected the pathways through which environmental variables cascade to the sizes of individual home ranges to be much more complex in large than small species (Haskell et al., 2002), but we did not find such evidence. In a recent paper, Noonan et al. (2020) provide evidence that investigating allometric relationship of the sizes of home ranges from conventional KDE method could be misleading as the sizes of home ranges tend to be increasingly underestimated as species body mass increases. This exemplifies that the endeavor of reviewing the variation in the sizes of home ranges requires first and foremost the development of a rigorous and accepted methodology for estimating the sizes of home ranges, while the conceptual framework is already well-anchored since the theoretical approaches developed in the late 1960s (McNab, 1963; Jetz et al., 2004).

Group Size

Unlike what we posited, the support for the biological hypotheses did not decrease in species with large group size. Our rough categories of group size at the species level and the low number of species included in our review might have prevented us from detecting the influence of group size on the relationship between resource availability or landscape heterogeneity and the sizes of home ranges. In most group-living species, the situation is complex (Prox and Farine, 2020) because groups vary in size (e.g., fission-fusion societies, Aureli et al., 2008). The lability of group sizes and also the varying degree of home range overlap among individuals of a given social unit (Mcloughlin et al., 2000) may however prevent detecting a resource-to-home range size relationship, when group size is not controlled.

Diet

The support to the resource hypothesis was higher for species with a “cattle-type” rumen (Godron and Clauss, 2010), which contradicted our expectation. This could be caused by the difficulty to obtain a reliable metric for resource availability.
in moose-type species (mostly browsers, Codron and Clauss, 2010), whose resources are difficult to evaluate using remote sensing. In addition, grass (consumed by individuals of species with cattle-type rumen to different degrees) produces most of its rich foliage during a limited time frame and quickly disappears when passing into the non-productive season, while browse such as ivy or bramble, due to their deep roots, can continue to produce some buds and foliage even after the end of the productive season (Jarman, 1974; Shipley, 1999). This difference in growth rates of resources can render cattle-type species to be more sensitive to resource availability. As broached earlier, a proper assessment of species-specific foodscape and of the search and exploitation balance in time budget and movements is required for a better understanding of how resource availability and landscape heterogeneity shape the sizes of home ranges in species with different diets (Shipley, 1999) and morpho-digestive constraints (Clauss et al., 2003).

Other Drivers of Variation in the Sizes of Individual Home Ranges

Individual Attributes

We only analyzed the effect of resource availability and landscape heterogeneity on the variation in the sizes of home ranges in relation to species-specific diet and group size. Nonetheless, other factors shape among-individual variation in the sizes of home ranges, such as individual attributes (e.g., sex, age, or body condition). As critical periods in terms of energetic expenditures differ between sexes (i.e., mating in males, late gestation-early lactation in females) and because females cannot move a lot after parturition, intraspecific variation in the sizes of home ranges emerges (e.g., Bowyer, 2004; Ruckstuhl, 2007). The size of an individual’s home range can also vary as a function of age (Tao et al., 2016), either because individuals gain more experience and a better knowledge of resource distribution when aging (Said et al., 2009) or because they are increasingly less efficient at exploiting a large area (Froy et al., 2018). More generally, intraspecific variation in the internal state of individuals could lead to contrasted costs of movement (McNab, 1963) and therefore to different responses of the sizes of home ranges to changes in environmental variables.

Population Density

We omitted population density as a possible driver of the size of individual home range, as we considered population density to be an emergent property of combined individual home ranges and overlap among individuals (Damuth, 1981). Even though the sizes of home ranges can decrease with density, as shown by Kjellander et al. (2004) on roe deer, changes in home range overlap could blur this relationship (Ietz et al., 2004). More generally, a detailed appraisal of the social system should be inseparable from the study of the home range-to-density relationship (Damuth, 1981; Mcloughlin et al., 2000).

Coexisting With Other Species

In addition, most individuals share their home ranges with members of sympatric species with overlapping niches, which leads to potential competing or facilitating relationships (Buchmann et al., 2013; du Toit and Olff, 2014). Individuals living in a multispecies environment, where resources are depleted at a fast rate, should have larger home ranges than individuals living alone in a given area (Buchmann et al., 2013). The inverse may happen when facilitation occurs (du Toit and Olff, 2014). The reason the community setting is seldom considered in single-species studies of variation in the sizes of home ranges may be the strong bias of studies from the northern hemisphere, where the research tradition is more single species based, in contrast to Africa where comprehensive species and community approaches are more common (e.g., starting in the early 1970s with the works of Estes, 1974). Ungulates are also prey to various predators in all ecosystems (Montgomery et al., 2019). The impact of natural predation and human hunting on the sizes of individual home ranges within a population remains to be evaluated, accounting for both ungulate life history traits (Hopcraft et al., 2010) and predator hunting modes (cursorial vs. ambush, Say-Sallaz et al., 2019).

Recommendations

We acknowledge that in the last 15 years, several authors have reviewed in depth the concept of home range and its estimation, providing important recommendations regarding the methodology and reporting of information in published studies (for recent studies, Noonan et al., 2019, 2020; Péron, 2019). More than 10 years ago, Laver and Kelly (2008) urged researchers to follow a unified methodology for estimating animal home ranges and recommended a minimal report of information for a better reproducibility and robust comparisons among studies. Likewise, we can only insist that it would be most useful, for comparative endeavors like ours, that authors provide estimates of the sizes of home ranges obtained with different methods, if only as Supplementary Information. In addition, previous reviews provided insightful information and critical perspectives regarding the definition of a home range and the estimation of its size (Börger et al., 2008; Fieberg and Börger, 2012; Péron, 2019). In a prospective review, Börger et al. (2008) advocated for the development of a general mechanistic model of animal home range behavior that would unify movement, individual resource requirements, and statistical models of home ranges. While this review did not directly provide guidelines on how to enhance the reporting and standardization of results of home range studies, it proposed a general approach, unifying our understanding of links between small-scale movements and home range behavior. We refer the readers to these previous reviews for discussion about the concept of home range, its definition and estimation, and the future avenues of research regarding the upscaling of movements to home ranges. In our systematic review, we encountered other hurdles, such as the methodological and biological inconsistencies regarding the independent variables and the statistical models. This led us to summarize some recommendations (Table 4) that should
TABLE 4 | Recommendations for authors that should help quantitative studies aiming at evaluating the effect of the environment and life history traits on the size of home range.

Recommendations

A. Study design and methodology related to estimating home ranges (see also, Laver and Kelly, 2008)
- Sample size is crucial: mark enough individuals;
- The choice of home range estimator is crucial and for comparative purposes, repeating the analyses with different estimators (to be provided as Appendix for example) would be helpful;
- State the bandwidth and the kernel used when making kernel home range estimates. Without both information, comparisons of kernel estimates of utilization distributions or estimates of sizes cannot be made. If, however, bandwidth and kernel are presented, then biases of estimates can be known and used when making comparisons.
- Evaluate the movement stationarity vs. time-window relationship to choose a meaningful temporal scale regarding (1) the definition of a home range, (2) the temporal and spatial scales at which environmental variables are estimated (see also, Börger et al., 2006b), and (3) the body mass of the species.

B. Identifying and measuring environmental variables
- The effect of food resource availability should not be studied without accounting for other complementary resources and landscape structure;
- Evaluate the species’ foodscape when selecting metrics of food resources;
- Consider the different components of landscape heterogeneity explicitly (i.e., fragmentation, diversity, complementation) and express specific hypotheses for each component;
- Consider reducing the number of metrics of landscape heterogeneity and formulating clear hypotheses on expected causal relationships.

C. General statistical considerations
- Scale variables to allow comparison across studies;
- Report standardized \( \beta \) estimates of models for a better comparative approach;
- Always report \( \beta \) estimates even in cases where the effect was not statistically significant;
- Report the transformation used on variables;
- Evaluate the log-log relationship between home range, body mass and biomass;
- Correct for multiple testing when multiple metrics are evaluated, and especially when these metrics are correlated (e.g., \textit{Fragstats} metrics);
- Express the expected relationship in formal mathematical terms;
- Take into consideration bias when measuring the metric value (e.g., Metric value = True value + bias + error).

D. Biological considerations
- Provide a clear definition of home range and of the rationale behind its studied properties (size, shape, temporal scale, spatial scale—see also, recommendations in A.) (see also, Börger et al., 2008; Péron, 2019);
- For social species, consider the sociality level (overlap with other conspecific, group size, group stability) for each individual (Péron, 2019);
- Evaluate the individual exposure to lethal (predation, hunting, collision) and to non-lethal (disturbances) risks at the individual level;
- For social species, consider the sociality level (overlap with other conspecific, group size, group stability) for each individual (Péron, 2019).
- Account for sympatric species of the same guild and overlapping trophic niche that could lead to modification of individual space use through competition (interference, exploitation) or facilitation processes.

DATA AVAILABILITY STATEMENT
The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

AUTHOR CONTRIBUTIONS
JS-F, KA, GY, and AL conceived this work and designed the methodology with inputs from J-MG, MG, and NM. JS-F and KA collected and analyzed the data with inputs from VR. JS-F, KA, GY, and AL led the initial writing of the manuscript. All authors contributed critically to the manuscript and gave final approval for publication.

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SUPPLEMENTARY MATERIAL

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fevo.2020.555429/full#supplementary-material
Vercauteren, K. C., and Hygnstrom, S. E. (1998). Effects of agricultural activities and hunting on home ranges of female white-tailed deer. J. Wildlife Manag. 62, 280–285. doi: 10.2307/3802889

Viana, D. S., Granados, J. E., Fandos, P., Pérez, J. M., Cano-Manuel, F. J., Burón, D., et al. (2018). Linking seasonal home range size with habitat selection and movement in a mountain ungulate. Mov. Ecol. 6, 1–11. doi: 10.1186/s40462-017-0119-8

Volampeno, M. S. N., Masters, J. C., and Downs, C. T. (2011). Home range size in the blue-eyed black lemur (Eulemur flavifrons): a comparison between dry and wet seasons. Mamm. Biol. 76, 157–164. doi: 10.1016/j.mambio.2010.04.005

Walter, W. D., VerCauteren, K. C., Campa, H., Clark, W. R., Fischer, J. W., Hygnstrom, S. E., et al. (2009). Regional assessment on influence of landscape configuration and connectivity on range size of white-tailed deer. Landscape Ecol. 24, 1405–1420. doi: 10.1007/s10980-009-9374-4

Widmer, O., Said, S., Miroir, J., Duncan, P., Gaillard, J.-M., and Klein, F. (2004). The effects of hurricane Lothar on habitat use of roe deer. Forest Ecol. Manag. 195, 237–242. doi: 10.1016/j.foreco.2004.02.021

Wiens, J. A., Crawford, C. S., and Gosz, J. R. (1985). Boundary dynamics: a conceptual framework for studying landscape ecosystems. Oikos 45, 421–427. doi: 10.2307/3565577

Wilson, R. P., Shepard, E. L. C., and Liebsch, N. (2008). Prying into the intimate details of animal lives: use of a daily diary on animals. Endang. Species Res. 4, 123–137. doi: 10.3354/esr00064

Conflict of Interest: The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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