Habitat selection of roe deer (Capreolus capreolus)

in a landscape of fear shaped

by human recreation

MASTER THESIS

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Abstract

Predation risk varies in space and time across the landscape and thus, causes a heterogeneous "landscape of fear". Within this landscape, roe deer (Capreolus capreolus) adapt their habitat selection to minimize the current risk and to maximize energy gain as well as energy saving. As roe deer perceive human disturbance as predation risk, human activities affect habitat selection of roe deer. This may lead to the avoidance of high quality habitats and therefore to a reduction in individual and population fitness. Nevertheless, the role of recreational activity as a human disturbance for roe deer is rarely investigated. We analysed how non-motorized recreational activity affects habitat selection of twelve GPS-collared roe deer in a peri-urban landscape of Switzerland. Recreational activity differs between weekend and workdays as well as between daytime, nighttime and twilight periods. Thus, we predicted that habitat selection of the roe deer varies among these time periods. We estimated individual home ranges (HR) and core areas (CA) by using kernel density estimation with the smoothing parameter $h_{ad hoc}$. We calculated HR and CA sizes for the different time periods as well as percent overlaps of these measures between weekend and workday, day and night, day and twilight as well as night and twilight. Further, we modelled habitat selection of each roe deer within their HR by using the negative binomial resource selection function (NB RSF). Home range and CA sizes as well as habitat selection did not differ between weekend and workday. In contrast, diurnal HR and CA were smaller than both nocturnal and twilight HR and CA, respectively. Roe deer generally avoided roads and used flat areas as well as open areas more frequently during the night than during the day, likely as a result of human disturbance. Further, roe deer selected habitats with high canopy cover during the night and forest stands with high conifer proportion in general. Additionally, habitat selection of roe deer exposed to intense recreational activity was similar to that of roe deer exposed to low recreational activity. Our findings indicate that roe deer adapt habitat selection over the course of a day as a result of recreational activity. However, weekly variation in recreational activity did not affect habitat selection of roe deer. Roe deer usually exhibit behavioural plasticity and adapt well to human disturbance. The studied roe deer also adapted to but did not tolerate recreational activity. Thus, human disturbance still influenced the behaviour of the roe deer and future studies should discuss the effect on individual and population fitness.

Keywords: behavioural adaptations, diel patterns, Global Positioning System (GPS), habitat selection, home range, human disturbance, kernel density estimation, landscape of fear, resource selection function, recreation, roe deer (*Capreolus capreolus*), spatial behaviour, weekly patterns

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Introduction

Spatially variable predation risk across a landscape has been termed a "landscape of fear" (Laundre et al. 2001). It depends on the characteristics of the particular predator (e.g. time of activity) and the ability of the prey to avoid the predation (Brown 1999, Frid et al. 2002). The hills and valleys of the landscape of fear are shaped by different levels of predation risk depending on the habitat type (Laundre et al. 2001). Because the prey intends to minimize predation risk, predators not only kill, but also influence habitat selection of their prey (Lima et al. 1990, Brown et al. 1999). In general, prey react in two different ways to the predation risk in the landscape of fear: moving from risky to safer habitats or enhancing vigilance (Brown 1999). Prey exhibit this responses both to animal predators (e.g. Laundre et al. 2001, Hernandez et al. 2005) and human activities (e.g. Sibbald et al. 2011, Ciuti et al. 2012). Thus, also human activity can cause a landscape of fear (Frid et al. 2002, Roesner et al. 2014) which may be different from those of natural predators (Lone et al. 2014) and may influence the behaviour of the prey even stronger than natural predators do (Ciuti et al. 2012).

Urban sprawl, habitat loss and fragmentation increase around the world (e.g. Antrop 2004, Jaeger et al. 2007, He et al. 2014), leading to greater human disturbance in peri-urban areas where recreation areas and wildlife habitats overlap (Herbold 1995, Hewison et al. 2001, Markovchick-Nicholls et al. 2008). Several negative short- and long-term effects of human disturbance on ungulates are documented (reviews: Vistnes et al. 2007, Stankowich 2008, Harris et al. 2014). Short-term effects comprise increased heart rate in roe deer (Capreolus capreolus) and red deer (Cervus elaphus; Reimoser 2012), enhanced vigilance in roe deer (Benhaiem et al. 2008) and elk (Cervus canadensis; Ciuti et al. 2012) as well as flight response in various ungulate species (Taylor et al. 2003, Stankowich 2008). Even if ungulate populations do not respond immediately or on a local scale to human disturbance, and thus seem to be habituated, they still can be affected in a long-term or on a broad scale (Vistnes et al. 2007, Bejder et al. 2009). Long-term effects include shift from crepuscular to nocturnal activity in Hainan Eld's deer (Cervus eldi hainanus; Pan et al. 2010), reduced diurnal activity in elk (Naylor et al. 2009) and changing habitat selection in various ungulate species, e.g. moose (Alces alces; Harris et al. 2014), mountain caribou (Rangifer tarandus caribou; Seip et al. 2007), red deer (Sibbald et al. 2011), elk (Preisler et al. 2005), sika deer (Cervus nippon; Uzal et al. 2013) and roe deer (Benhaiem et al. 2008). Enhanced vigilance (Benhaiem et al. 2008, Ciuti et al. 2012), flight response (Colman et al. 2012) as well as the alteration of habitat selection (Hernandez et al. 2005, Benhaiem et al. 2008) may reduce the intake of food and especially of high quality food. Accordingly, disturbance may lead to a negative energy budget due to reduced food intake or enhanced energy expenditure (Houston et al. 2012) and hence may reduce population fitness (Ciuti et al. 2012). For instance, predation risk of human (Phillips et al. 2000) as well as wolves (*Canis lupus*; Creel et al. 2007) are reported to reduce calving rate of elks. Further, Nilsen et al. (2004) reported that the quality of the winter habitat is of great importance for the litter size of roe deer in the subsequent spring. Pettorelli et al. (2001) found that population density and fawn body weights of roe deer increases with increasing habitat quality. Thus, the knowledge about the landscape of fear and its consequences on the prey is crucial for wildlife conservation (Markovchick-Nicholls et al. 2008) and for understanding population dynamics (Laundre et al. 2014).

Roe deer prefer forest habitats (Mysterud et al. 1999, Nilsen et al. 2004), but also inhabit meadows, pasture and moorland if forest is sparse or disturbance is low (Hewison et al. 2001, Bonnot et al. 2013). Although they may occupy various forest types (e.g. Ratikainen et al. 2007, Pellerin et al. 2010, Ewald et al. 2014), they prefer forest stands with low visibility (Tufto et al. 1996, Borkowski et al. 2008), high food supply (Tufto et al. 1996, Pellerin et al. 2010) and high food quality (Moser et al. 2006, Pellerin et al. 2010). In winter, they prefer conifer forest (Mysterud et al. 1995, Mysterud et al. 1999, Ratikainen et al. 2007), mature stands (Herbold 1995, Mysterud et al. 1999) and high canopy cover (Ratikainen et al. 2007). Resting roe deer prefer higher canopy cover than active roe deer (Ewald et al. 2014). Roe deer generally avoid roads (Jiang et al. 2009), especially during the day (Imfeld 1996, Bonnot et al. 2013), in landscapes with sparse woodland (Bonnot et al. 2013) and near buildings (Coulon et al. 2008). The avoidance of settlements is higher in open areas than in forest, higher in landscapes with less than with vast woodland (Coulon et al. 2008) and higher during the day than during the night (Mysterud et al. 1999). Roe deer seek forest stands with denser concealment cover when disturbance is high than when it is low (Herbold 1995, Mysterud et al. 1999). Further, roe deer may retreat to steep slopes during the day (Imfeld 1996) and enhance activity during the night (Guthorl 1994), both as a consequence of avoiding human. Thus, roe deer exhibit behavioural plasticity and may adapt to human disturbance (Jeppesen 1987, Hewison et al. 2001, Bonnot et al. 2013), potentially better than other ungulates (e.g. red deer: Jiang et al. 2008; moose: Jiang et al. 2010). However, human infrastructures and disturbance still influence the behaviour and habitat selection of roe deer (e.g. Guthorl 1994, Jiang et al. 2009).

Most previous studies on habitat selection of roe deer were conducted in rural areas and focused on nonhuman variables (e.g. Ratikainen et al. 2007, Ewald et al. 2014) or on human infrastructures (e.g. Coulon et al. 2008, Jiang et al. 2009, Bonnot et al. 2013). Thus, little is known about the influence of the temporal variation in recreational activities on habitat selection of roe deer in a peri-urban area (but see Herbold 1995, Imfeld 1996). We filled this gap by investigating how roe deer use a landscape of fear shaped by mostly nonmotorized recreational activities in a peri-urban region of Switzerland. Therefore, we analysed home range (HR) and core area (CA) sizes, HR and CA locations as well as habitat selection of twelve roe deer in the winter 2013/2014. As the recreational activity in the peri-urban area of Zurich exhibits diel and weekly variations (Wyttenbach 2012), we investigated how these temporal as well as regional variations affect habitat use of roe deer. We predicted that HR and CA sizes differ between weekends (WE) and workdays (WD) as well as among daytime, nighttime and twilight periods. Further, we predicted that the habitat selection differ between WE and WD as well as between day and night, regarding the use of open and steep areas as well as the use of areas near roads. Additionally, we predicted that regional variation in recreational use affects HR and CA size as well as the habitat selection.

Methods

Study site

The study was conducted between 16.10.2013 and 15.04.2014 in the Sihltal area south of Zurich, Switzerland (N 47° 15'; E 8° 33'). The study area covers 3646 ha, encompassing forest (2064 ha, 57%), meadows (1415 ha, 39%) as well as settlements and waters (167 ha, 4%), at elevations ranging from 500 to 900 m above sea level (figure 1). Forests are mainly dominated by European beech (Fagus sylvatica) and partially by Norway spruce (Picea abies). European silver fir (Abies alba), Scots pine (Pinus sylvestris), European ash (Fraxinus excelsior) and sycamore maple (Acer pseudoplatanus) occur irregularly (ALN 2013). The nearest weather station (Wädenswil; 485 m above sea level) recorded long-term means for annual temperature and precipitation of 9.5 °C and 1390 mm, respectively (MeteoSwiss 2013). During the study period average temperature was 4.7 °C and snow height reached 0.2 cm on average and 5 cm at most (MeteoSwiss 2014). In the study region, no evidence exists for the occurrence of large carnivores such as lynx (Lynx lynx) or wolves for at least 120 years (KORA 2012). Hunting of roe deer males is permitted between 2nd May and 31st December and those of females and fawns between 1st September and 31st December (Article 19, Jagdverordnung). Most of the study site is situated within the agglomeration of Zurich and highly used by non-motorized recreational activities such as hiking, biking, jogging and dog walking. Other parts are situated in more rural and steeper areas with a sparse road system. The HR of the studied roe deer comprised 1.8 \pm 1.6 km/km² (mean \pm SD) surfaced roads as well as 10.1 \pm 3.7 km/km² gravel roads and gravel paths many of which are closed to public traffic.



Figure 1. Study area (black) and its location in Switzerland. The home ranges of the twelve roe deer (1 to 12) are outlined in red. Background map: © Federal Office of Topography, swisstopo.

Telemetry data and study animals

Between September and December 2013, twelve roe deer were caught using drive netting (Lopez-Olvera et al. 2009). After the application of a sedative (0.3 ml Azepromazine), the roe deer were collared with a GPS-GSM transmitter equipped with integrated activity sensors (Model GPS PLUS, VECTRONIC Aerospace, Berlin, Germany). Locations were obtained via GPS (Global Positioning System) and send as SMS (Short Message Service) via GSM (Global System for Mobile Communication) to the researcher. The transmitters were programmed to obtain a location every three hours (MET: 00:00, 03:00, ..., 21:00). Data from the first three days after capture were deleted because handling might have influenced initial behaviour (Morellet et al. 2009). Location gaps occurred if the GPS logger was not able to connect to the satellite within two minutes or to the GSM over several days. We used only GPS Data with a dilution of precision (DOP) value lower than ten, leading to a mean \pm SD data gap of 3.7 \pm 4.9% (table 1). The GPS accuracy was 5 m. All roe deer were adults and in good body condition. According to local expert classification, five and seven roe deer inhabit areas with a high and low intensity of recreational use, respectively (further referred as peri-urban and rural roe deer; table 1). The HR of the peri-urban roe deer exhibited a significant higher density of surfaced roads

(W = 32, P = 0.022) as well as lower distances to roads (W = 5, P = 0.048) than the HR of rural roe deer. We defined the following GPS samples:

- winter: all valid GPS locations,
- WE: locations from Saturday and Sunday,
- WD: locations from Tuesday and Thursday,
- locations at daytime (day, diurnal): 1h after sunrise to 1h before sunset,
- locations at nighttime (night, nocturnal): 2h after sunset to 2h before sunrise, and
- twilight: locations between day- and nighttime.

Table 1. Overview of the data obtained from twelve collared roe deer, four males (m) and eight females (f). Five individuals were exposed to intense and seven individuals to low recreational activity, living in periurban and rural areas, respectively. The number of valid GPS locations varied according to different sampling periods and percentages of missing locations.

Animal ID	Sex	Classification according to the intensity of recreational use	Valid GPS locations	Data gap	Sampling period
RD01	m	peri-urban	1400	4.0%	16.10.2013 - 15.04.2014
RD02	f	peri-urban	1423	2.3%	16.10.2013 - 15.04.2014
RD03	m	rural	1432	1.7%	16.10.2013 - 15.04.2014
RD04	f	rural	1411	3.2%	16.10.2013 - 15.04.2014
RD05	f	rural	1434	1.5%	16.10.2013 - 15.04.2014
RD06	m	rural	1223	19.1%	16.10.2013 - 15.04.2014
RD07	f	peri-urban	1402	3.0%	17.10.2013 - 15.04.2014
RD08	m	rural	1404	2.3%	18.10.2013 - 15.04.2014
RD09	f	rural	1238	1.9%	09.11.2013 - 15.04.2014
RD10	f	peri-urban	1192	1.0%	16.11.2013 - 15.04.2014
RD11	f	rural	1173	2.0%	17.11.2013 - 15.04.2014
RD12	f	peri-urban	1112	2.5%	24.11.2013 - 15.04.2014

Home range and core area estimations

The HR encompasses the area occupied by an individual at 95% probability during a specific period (Kernohan et al. 2001). In equivalence, the CA is usually defined as the 50% probability of occurrence (Laver et al. 2008). Various methods for estimating HR and CA exist (White et al. 1990), but kernel density estimation (KDE) is, besides the minimum convex polygon (MCP), the most common (Laver et al. 2008). The KDE is often preferred over other methods due to several advantages such as i) high accuracy (Boerger et al. 2006, Pebsworth et al. 2012), ii) less bias due to temporal autocorrelation (Swihart et al. 1997), iii) utility distribution as basis (Kernohan et al. 2001, Gitzen et al. 2006), iv) identification of multiple centres of activity (Worton 1987, Pebsworth et al. 2012) and v) exclusion of holes (Lichti et al. 2011; but see Getz et al. 2004). In KDE, a probability density function (kernel) is placed over each GPS location (Worton 1989). The overlapping density functions produce a gridded utility distribution (UD). The more locations are in and around a grid unit, the higher the value of this unit (Worton 1989, Seaman et al. 1996). How much a nearby location contributes to the value of the unit depends on the smoothing parameter (h; Silverman 1986, Worton 1989, Horne et al. 2006). Commonly, the smoothing parameter is chosen by least-squares cross-validation (h_{LSCV}; Laver et al. 2008). However, the appropriate smoothing parameter depends on the sample size, the isopleth, the spatial spread as well as the pattern of the data and the research question (e.g. Blundell et al. 2001, Boerger et al. 2006, Gitzen et al. 2006, Lichti et al. 2011). The extent of the HR and CA is received from the UD by applying the specific isopleth (e.g. 95%; White et al. 1990).

We estimated HR and CA sizes using the package adehabitatHR (Calenge 2011) in the software R 3.1.1 (R Core Team 2014) for the winter, WE, WD, day, night and twilight samples. We used a common bivariate normal kernel (Gaussian; Laver et al. 2008) with a grid size of 25 x 25 meters as it resulted in reasonable areas. Further, we considered a motorway on the north-east HR border of RD01, RD07 and RD12 as a non-traversable barrier. The winter sample and the five subsamples (WE, WD, diurnal, nocturnal, twilight) had a mean \pm SD sample size of 1320 \pm 121 (table 1) and 353 \pm 43 locations (Appendix I), respectively, which is above the recommended minimum (Kernohan et al. 2001, Boerger et al. 2006, Pellerin et al. 2008). The smoothing parameter was fixed for all locations (fixed kernel; Seaman et al. 1999). The h_{LSCV} used resulted in fragmented HR and CA, an outcome that was also found unsuitable for estimating roe deer HR by Pellerin et al. (2008). Therefore, we used the ad hoc technique (h_{ad hoc}) proposed by several authors (e.g. Boerger et al. 2006, Pebsworth et al. 2012, Schuler et al. 2014). The h_{ad hoc} allowed to test different proportions of the reference smoothing parameter (h_{ref}) until a continuous boundary was reached (Pebsworth et al. 2012). The reference smoothing parameter is defined as h_{ref} = 0.5 * $n^{0.167}$ * (*SD*(*x*) + *SD*(*y*)), where *n* equals the number

of locations and *SD(.)* is the standard deviation each of the *x* and the *y* coordinates (Seaman et al. 1996). As the smoothing parameter depends on the sample size (Gitzen et al. 2006), we randomly drew two thirds of the original day and night samples for the HR and CA analysis. Thereby, we ensured similar sample size for the five subsamples (WE, WD, diurnal, nocturnal, twilight) and thus could apply the same $h_{ad hoc}$ for all of them. For winter HR and CA $h_{ref} * 1$ gave a continuous boundary for most of the twelve roe deer and thus $h_{ad hoc}$ equalled h_{ref} . We reduced the proportion to $h_{ad hoc} = h_{ref} * 0.9$ for the five subsamples to get similar buffers around the GPS locations like in the winter HR and CA. Because the absolute values of the smoothing parameters depend on the spatial pattern of the locations (Lichti et al. 2011), they were not consequently reduced in the subsamples (Appendix I), leading to higher biases in the UD of winter HR and CA than in the UD of the five subsamples (Fieberg 2007). However for our study, the UD was not important as we investigated HR and CA sizes as well as percent overlap which base on the shapes of the HR and CA. We calculated percent overlap each of the HR and the CA between the following subsamples:

- weekend and workday (WE/WD),
 day and twilight, as well as
- day and night,
 night and twilight.

The percent overlap is defined as $HR_{i, j} = A_{i, j} / A_i$ (Kernohan et al. 2001). In our study, $A_{i, j}$ is the overlap between the subsamples *i* and *j* and A_i is the area of the subsample *i*. The mean percent overlap between subsample *i* and *j* is the mean of $HR_{i, j}$ and $HR_{j, i}$ with the subsample *i* and *j* as the divisor, respectively. Other methods for measuring overlap exist (Fieberg et al. 2005), but percent overlap is the most intuitive and quite widely used (Kernohan et al. 2001), also in recent literature (e.g. Richard et al. 2014). However, it provides no information about the use within the overlap because the UD is ignored (Fieberg et al. 2005).

Habitat selection

Roe deer do not move randomly through the landscape and within their HR (e.g. Coulon et al. 2008, Bonnot et al. 2013, Ewald et al. 2014). They use habitats selectively, which means disproportionately to their availability (Johnson 1980). If GPS locations are the data basis, habitat selection is often assessed with a resource selection function (RSF; Leban et al. 2001, Manly et al. 2002, Northrup et al. 2013), also called habitat selection function (HSF; Aarts et al. 2013). Habitats are patches with a certain combination of environmental conditions, including biotic and abiotic factors (Aarts et al. 2008). Habitat selection is the association of an animal with different habitat types (Silvy 2012). Alternatively to "habitat selection", some literature uses the term "habitat preference" (e.g. Beyer et al. 2010, Aarts et al. 2013). Traditionally, preference refers to the use of habitats if they are offered on an equal basis and selection is the process behind prefer-

ence (Johnson 1980). As habitats are usually not offered on an equal basis in nature, preference can also be defined relative to the availability of habitats (Aarts et al. 2008). We assessed habitat selection depending on diel, weekly and regional variation in recreational activity and thus investigated a process behind habitat preference. Regarding RSF, two further terms are important: use and availability. In GPS studies, the use is represented by the quantity of the GPS locations in each habitat type during the study period (Johnson 1980). The availability, defined as the quantity of habitats accessible to the animal during the study period (Johnson 1980), has to be defined by the researcher. The dimension of availability is crucial as it influences the covariates of the RSF (Aarts et al. 2008, Beyer et al. 2010, Aarts et al. 2013, Northrup et al. 2013).

The RSF include a variety of methods such as selection indices, compositional analysis, logistic regression and discrete choice models (Manly et al. 2002, Thomas et al. 2006, Silvy 2012). Emerging methods include different kinds of machine learning techniques such as maximum entropy (McDonald et al. 2013). A RSF results in a value for each resource – here habitat type – which is proportional to the probability of being used by an organism (Manly et al. 2002). Three designs to fit a RSF exist (see Boyce et al. 2002, Aarts et al. 2012). If no absence data are available, the use-availability design with logistic regression is the most common method (Northrup et al. 2013). Nielson et al. (2013) proposed a simple count design for GPS data: a negative binomial resource selection function (NB RSF). This approach is based on the log-negative binomial regression (NB2; Hilbe 2011, Nielson et al. 2013). The application of the NB RSF in Sawyer et al. (2009) was identified as state-of-the-art analysis for GPS data by Silvy (2012). It has several advantages over other methods such as i) the modelling of an intensity rather than "used" and "unused", ii) no bias due to temporal autocorrelation and iii) a simple implementation (Nielson et al. 2013).

Modelling habitat selection with the NB RSF approach

We fitted WE, WD, day and night GPS locations of each roe deer to habitat variables using the NB RSF approach (Nielson et al. 2013). Based on these models, we analysed habitat selection in general as well as differences between WE and WD and between day and night. The modelling consisted of four steps: (1) We prepared habitat data. (2) We calculated the number of GPS locations and the values of the habitat variables in defined sample units. The number of locations was calculated separately for the WE, WD, day and night subsample. (3) We modelled the number of GPS locations within the sample units with the NB RSF approach (Nielson et al. 2013). Models were calculated using locations from either the WE, WD, day or night subsample, leading to four different models for each roe deer. (4) We analysed selection patterns and tested for differences between WE and WD as well as between diurnal and nocturnal models, using stat-

istics over all twelve roe deer. The steps (2) to (4) were conducted within the Software R 3.1.1 (R Core Team 2014).

(1) We prepared habitat data in ArcMap 10.1 (ESRI, Redlands, CA, USA). We extracted canopy cover (*canopy*), conifer proportion (*conifers*) and forest edge from forest maps which are based on infrared images from 2010 (ALN 2013). *Canopy* and *conifers* had a value of zero outside the forest. Motorways, settlements, buildings, ponds and rivers (geodata swisstopo; DV084370) were merged to the layer *unavailable*. We created rasters with a resolution of 5 x 5 m for *canopy, conifers* and *unavailable*. Further, we generated similar rasters consisting of the distance to the roads (*dist_road*), to the buildings or settlements (*dist_build*) and to the forest edges (*dist_forest_edge*), with the tool euclidean distance. In addition, *slope* was calculated based on a digital elevation model (DHM25; geodata swisstopo; DV084370) resulting in a raster with resolution 25 x 25 m. All data had a positional accuracy of at least 10 m (ALN 2013, swisstopo 2007).

(2) In the NB RSF approach, the count of GPS locations within a defined sampling unit is fitted to the habitat data of the unit:

(Eq. 1)
$$\ln [E(r_i)] = \ln (total r) + \beta_0 + \beta_1 X_1 + ... \beta_p X_p$$

where r_i is the absolute number of GPS location in the sampling unit *i*, *total r* is the total number of GPS locations, β_0 is the intercept, β_1 to β_p are unknown coefficients for the model variables X_1 to X_p and E(.) denotes the expected value (Nielson et al. 2013). Thus, the response variable is the absolute count of GPS locations per sampling unit. However, as we included the total number of GPS locations in the offset term, ln (*total r*), we rather modelled the relative count per sampling unit (Sawyer et al. 2006, Sawyer et al. 2009). As the sum of r_i across all sampling units should not be guaranteed to equal *total r* (Nielson et al. 2013), we

generated a systematic circular sampling with circle diameter 25 m which allowed interspace (figure 2). We intended to analyse habitat selection within the individual HR of the roe deer (third order selection; Johnson 1980) and thus, deleted circles whose centre lay outside the HR. Thereby, we also deleted GPS locations which neither lay within a circle nor within the HR. Accordingly, *total r* in our study was the sum of GPS locations which either lay within the HR of the roe deer



Figure 2. Circular sampling design; circles had a diameter of 25 m and were put inside each home range.

or within a circle whose centre lay within the HR. In the choice of the circle diameter, we considered the spatial error of the telemetry and the habitat data, the proportion of circle units with zero locations (mean ± SD: 70.4 \pm 1.5%), the sample size for each roe deer (647 \pm 409; Appendix III), the acquisition of the use pattern of the roe deer as well as the match with the spatial heterogeneity of the landscape. Thereafter, we calculated the absolute number of GPS locations within each circle using the WE, WD, diurnal and nocturnal data set separately. The mean number of locations per circle was 0.73 ± 0.14 and on average, $77.8 \pm 0.1\%$ of all locations lay within a circle (Appendix III). The environmental data for each circle were calculated with the weighted mean of the raster values for continuous variables (canopy, conifers, slope, dist_road, dist build, dist forest edge) or equated with the value capturing the highest area within the circle for the categorical variable unavailable. We deleted all circles with the attribute unavailable = 1, because they included areas that are not occupied but in the HR of some roe deer due to trade-offs in the selection of the smoothing parameter h_{ad hoc}. To check the presence of collinearity between habitat variables, we calculated Spearman rank correlation coefficients (r_{spearman}) of all possible pairwise combinations. The r_{spearman} between canopy and conifers was calculated by using only forested circles, as these habitat variables exceeded zero, and thus contributed to the model, only in this circles. In doing so, the Spearman rank correlation coefficient did not exceed $|r_{spearman}| > 0.60$ for any combination of habitat variables. As we did not intend to rank the influence of the different habitat variables, but rather compared the effect of one habitat variable among the roe deer, we did not standardize the habitat variables (Urban et al. 2011). With this sampling design, we assessed the use (number of locations per circle) and the availability (circular units within the HR) for each roe deer individually (design III study; Thomas et al. 1990, Manly et al. 2002).

(3) First, we identified the most important model variables. Therefore, we pooled WE and WD locations across all roe deer. Thus, r_i in Eq. 1 was the absolute number of WE and WD locations in the circle unit *i* and *total r* the sum of all WE and WD locations which lay within the HR of any roe deer or within a circle whose centre lay within the HR of any roe deer. We fitted this use data with the function glm.nb in the package MASS (Venables et al. 2002) to the habitat variables using the linear and quadratic terms of all habitat variables as well as several interaction terms (table 2). We used automated stepwise selection in both directions on the basis of the Bayesian information criterion (BIC) to get the model formula with the most important model variables. We did the same procedure for the pooled day and night locations. The resulting model formula to fit the four samples WE, WD, day and night separately and for each roe deer individually to the habitat variables. Thus, r_i in Eq. 1 was the absolute number of either WE, WD, day or night locations in the circle unit *i* and *total r* the sum of either all WE, all WD, all day or all night locations of

one roe deer which lay within the HR or within a circle whose centre lay within the HR of the particular roe deer. Thus, we got twelve WE, WD, day and night models each and four models per roe deer.

(4) We calculated the odds ratios (odds ratio = $\exp(\beta_p) - 1$; Nielson et al. 2013) of the coefficients and compared them between WE and WD as well as between day and night models. Thus, we compared two models of the same roe deer each which resulted in a sample size of twelve for the statistical analysis regarding diel and weekly patterns. These analyses were independent of the availability definition as we compared models that base on the same availability (Beyer et al. 2010, Pellerin et al. 2010, Aarts et al. 2013). Further, we tested if the odds ratios of WE, WD, day and night in each case were significant higher or lower than zero, which would indicate a selection or avoidance of the model variable, respectively (Marzluff et al. 2004). We also compared the models between peri-urban and rural roe deer. In contrast to the first, these analyses may have been biased as the available habitat types differed among the roe deer (Mysterud et al. 1998). Further, we estimated an overall model each for WE, WD, day and night by averaging the coefficients of the individual roe deer models. The standard errors of the corresponding coefficients were also calculated from the roe deer models by leaving out the variation within the individual models (two stage approach; Marzluff et al. 2004, Fieberg et al. 2010). As an additional analysis, we calculated the percentage of GPS locations that lay within the forest using all locations (winter) or only a subset (WE, WD, day, night). We compared these measures between WE and WD, day and night as well as peri-urban and rural roe deer.

Table 2. Six linear habitat variables, their quadratic form as well as several interactions were included in the basis model. With automated stepwise selection the bold variables remained in the models for weekend / workday as well as for day and night. Canopy: canopy cover of tree crowns [%]; conifers: conifer proportion [%]; slope [°]; dist_road: distance to the nearest road [m]; dist_build: distance to the nearest building [m], dist_forest_edge: distance to the nearest forest edge [m].

Model variab	les
linear	canopy, conifers, slope, dist_road, dist_build, dist_forest_edge
quadratic	canopy ² , conifers ² , slope ² , dist_road ² , dist_build ² , dist_forest_edge ²
interactions	<pre>canopy : conifers, canopy : conifers², canopy : dist_road, canopy : dist_road², canopy :</pre>
	dist_build, canopy : dist_build ² , canopy : dist_forest_edge, canopy : dist_forest_edge ²

Significance testing

We used Wilcoxon signed rank test to analyse the difference between two paired samples such as WE and WD HR sizes or diurnal and nocturnal odds ratios. Friedman test and Nemenyi post-hoc test within the R package PMCMR (Pohlert 2014) were used to analyse differences among several paired samples such as diurnal, nocturnal and twilight HR sizes. Further, we used Wilcoxon rank sum test to analyse if odds ratios differed from zero and if measurements on peri-urban differed from those of rural roe deer. To analyse if a significant effect was stronger in peri-urban than in rural roe deer, we applied the Wilcoxon rank sum test on the differences of the results. For example, if diurnal HR were significantly smaller than nocturnal HR, we subtracted the diurnal from the nocturnal HR size and analysed this differences for peri-urban and rural roe deer using the Wilcoxon rank sum test. If the difference, obtained from the subtraction, was higher for the peri-urban than the rural roe deer, we concluded that the significant effect was stronger for the peri-urban than for the rural roe deer. All tests were conducted on a significance level of 0.05.

Results

Home range and core area estimations

Winter HR and CA sizes ranged from 13.3 to 107.6 ha (mean \pm SD : 44.5 \pm 30.9 ha) and from 3.3 to 21.8 ha (8.8 \pm 6.0 ha), respectively (figure 3, Appendix I). Home range and CA sizes did not differ between WE and WD (HR: V = 29, P = 0.470; CA: V = 30, P = 0.519). However, there was significant diel variation (HR: χ^2 = 17.17, df = 2, P < 0.001; CA: χ^2 = 12.67, df = 2, P = 0.002). The Nemenyi post-hoc test revealed smaller HR and CA for the diurnal sample compared to the nocturnal (HR: P = 0.022; CA: P = 0.012) and to the twilight sample (HR: P < 0.001; CA: P = 0.003). Further, the percent overlaps (figure 4 & 5; Appendix I) differed among WE/WD, day/night, day/twilight and night/twilight for the HR and the CA (χ^2 = 62, df = 7, P < 0.001). Nemenyi post-hoc test revealed that the HR overlaps usually did not differ from the corresponding CA overlaps. An exception was the subsample night/twilight where the CA overlap was smaller than the HR overlap (P < 0.001). In contrast, the CA overlap between day and night was not significantly smaller than the HR overlaps day/night were smaller than the overlaps WE/WD (HR: P = 0.019; CA: P < 0.001), that the HR overlaps day/night was smaller than the HR overlaps WE/WD (HR: P = 0.026), and that the CA overlap night/twilight was smaller than the CA overlap night/twilight (P = 0.026), and that the CA overlap night/twilight was smaller than the CA overlap wee peri-urban and rural roe deer (8 < W < 27; P > 0.149).



Figure 3. Home range and core area sizes calculated for the twelve roe deer using all points (winter) or only a subset. Weekend: Saturday and Sunday; workday: Tuesday and Thursday; day: 1h after sunrise to 1h before sunset; night: 2h after sunset to 2h before sunrise; twilight: time period between day and night. Different letters indicate significant difference at P < 0.05.



Figure 4. Overlap of home ranges and core areas; overlaps were calculated between the subsamples: weekend and workday, day and night, day and twilight as well as night and twilight. Weekend: Saturday and Sunday; workday: Tuesday and Thursday; day: 1h after sunrise to 1h before sunset; night: 2h after sunset to 2h before sunrise; twilight: time period between day and night.



Figure 5. Individual home ranges as well as diurnal and nocturnal core areas; A: RD02, RD03, RD06, RD08; B: RD01, RD10, RD12; C: RD07; D: RD05; E: RD04, RD09, RD11. Background map: © Federal Office of Topography, swisstopo (DV084370).

Habitat selection

We fitted individual models for twelve roe deer using GPS locations each from WE, WD, day and night. After the automated stepwise selection, all linear terms of the habitat variables, except *dist_build*, remained in the model (table 2). The statistical tests regarding habitat selection were conducted among the roe deer and thus had a sample size of twelve. The mean \pm SD dispersion of all individual models was 0.70 \pm 0.26, indicating minor underdispersion. Thus, the real number of GPS locations within a circle unit, had less variation than predicted by the model. However, according to the Vuong test the NB regression still performed better than the Poisson, the hurdle and the zero-inflated model (P < 0.001). McFadden's pseudo R² of all individual models was 0.32 \pm 0.15, and thus, the model explained on average 32% of the variation of the sampling units use. McFadden's pseudo R² did not differ between the individual WE and WD models (V = 37, P = 0.91), but was higher for the individual diurnal than the individual nocturnal models (V = 0, P = 0.002). Thus, the model might reproduce the diurnal habitat selection of the roe deer better than the nocturnal. The model variables of the individual HR were on average: *canopy*: 55 \pm 11%; *conifers*: 25 \pm 18%; *slope*: 12 \pm 6 °; *dist_road*: 39 \pm 17 m; *dist_forest_edge*: 73 \pm 35 m. The median value of *slope* and *dist_road* was 7 ° and 16 m, respectively. Thus, the availability of the habitat types differed among the roe deer.

Roe deer selected habitats with high canopy cover and a medium distance to road at WE and on WD (canopy WE: V = 66, P = 0.034; canopy WD: V = 71, P = 0.009; dist road WE & WD: V = 78, P < 0.001; $dist_road^2$ WE: V = 2, P = 0.001; $dist_road^2$ WD: V = 1, P < 0.001; table 3). Further, roe deer avoided habitats with hight canopy cover far away from roads on WD (V = 12, P = 0.034). While habitat selection did not differ between WE and WD (for all model variables: 26 < V < 54, P > 0.266), roe deer adapted habitat selection over the course of a day (table 4). Roe deer selected habitats with lower slope during the night than during the day (V = 8, P = 0.012; figure 6). The avoidance of areas with high canopy cover far away from roads, which also meant the selection of areas with low canopy cover near roads, was stronger during the night than during the day (V = 7, P = 0.009). Further, we found a tendency that the selection for high conifer proportion was weaker during the night than during the day (V = 14, P = 0.052). In general, roe deer selected habitats with high conifer proportion and medium distance to road, but avoided habitats which had both, a high conifer proportion and a high canopy cover (day *conifers*: V = 78, P < 0.001; night *conifers*: V = 68, P = 0.021; day *dist_road*: V = 75, P = 0.002; night *dist_road*: V = 78, P < 0.001; day *dist_road*²: V = 0, P < 0.001; night *dist_road*²: V = 5, P = 0.005; day *canopy : conifers*: V = 10, P = 0.021; night *canopy : conifers*: V = 11, P = 0.027; figure 7). Additionally, roe deer selected habitats with high canopy cover during the night (V = 71, P = 0.009), which gave a complex selection pattern with peaks in habitats with either very high canopy cover and very high conifer proportion, or low canopy cover and low conifer proportion (figure 8). Further, roe deer avoided habitats with high canopy cover if far away from roads during the night and thus, selected habitats with low canopy cover near roads (V = 5, P = 0.005; figure 8). Despite the models, we analysed habitat selection by calculating the percentage of GPS locations that was situated within the forest. Roe deer spent $84 \pm 12\%$ of their time within the forest when using all GPS locations. Nine and four of the twelve roe deer spent more than 75% and 90% of their time, respectively, within the forest using all GPS locations. While the use of forest areas did not differ between WE and WD (V = 40, P = 0.970), roe deer spent more time within the forest during the day (96 \pm 4%) than during the night (73 \pm 18%; V = 78, P < 0.001; figure 9). During the day, all roe deer spent more than 75%, and eleven of twelve roe deer spent more than 90% of their time within the forest. During the night, six and three of the twelve roe deer spent more than 75% and 90% of their time, respectively, within the forest.

Table 3. Odds ratios of the overall WE and WD models. These odds ratios are equal to the mean \pm SE of the odds ratios from the individual WE and WD roe deer models. The stars behind the odds ratios indicate on which significance level the odds ratios were different from zero: * P < 0.05, ** P < 0.01, *** P < 0.001. The column WE/WD contains the significance level on which a model variable differed between the WE and the WD models. No model variable differed significantly between the WE and the WD models.

Model variable	Mean ± SE of the odds ratios of individual WE models	Mean ± SE of the odds ratios of individual WD models	WE / WD
Intercept	- 0.999820 ± 5.257612	- 0.999888 ± 15.095409	
canopy [%]	$0.011515 \pm 0.011622 *$	0.018786 ± 0.026347 **	
conifers [%]	0.069396 ± 0.191034	0.065524 ± 0.198818	
conifers ²	0.000478 ± 0.007886	0.002495 ± 0.014422	
slope [°]	$- 0.008031 \pm 0.084087$	- 0.021754 ± 0.087564	
dist_road [m]	$0.047550 \pm 0.032154 ***$	0.052211 ± 0.040020 ***	
dist_road ²	- 0.000300 ± 0.000261 **	- 0.000389 ± 0.000584 ***	
dist_forest_edge [m]	- 0.002157 ± 0.056869	0.009016 ± 0.067857	
dist_forest_edge ²	0.000018 ± 0.000336	$- 0.000015 \pm 0.000376$	
canopy : conifers	- 0.000677 ± 0.002201	- 0.000666 ± 0.002406	
canopy : conifers ²	$- 0.000005 \pm 0.000093$	- 0.000031 ± 0.000173	
canopy : dist_road	- 0.000113 ± 0.000157	- 0.000105 ± 0.000218 *	
canopy : dist_forest_edge	0.000179 ± 0.000643	0.000059 ± 0.000772	
canopy : dist_forest_edge ²	-0.000002 ± 0.000005	- 0.000002 ± 0.000006	

Table 4. Odds ratios of the overall day and night models. These odds ratios are equal to the mean \pm SE of the odds ratios from the individual day and night roe deer models. The stars behind the odds ratios indicate on which significance level the odds ratios were different from zero: * P < 0.05, ** P < 0.01, *** P < 0.001. The column day/night contains the significance level on which a model variable differed between the day and the night models.

Model variable	Mean ± SE of the individual day m	e odds ratios of odels	Mean ± SE of individua	Day / night	
Intercept	- 0.999990 ± 1	07.951119	- 0.999520	± 4.854528	
canopy [%]	0.020942 ±	0.050699	0.010895	± 0.012436 **	
conifers [%]	0.312428 ±	0.264294 ***	0.089495	± 0.131387 *	
conifers ²	- 0.002554 ±	0.006476	- 0.001748	± 0.003741	
slope [°]	0.020891 ±	0.092484	- 0.028637	± 0.076161	*
dist_road [m]	$0.058095 \pm$	0.047818 **	0.043486	± 0.030042 ***	
dist_road ²	- $0.000519 \pm$	0.000509 ***	- 0.000334	± 0.000446 **	
dist_forest_edge [m]	0.002449 ±	0.151097	- 0.011068	± 0.038336	
dist_forest_edge ²	- 0.000146 ±	0.000916	0.000098	± 0.000228	
canopy : conifers	- 0.002742 ±	0.003405 *	- 0.000845	± 0.001280 *	
canopy : conifers ²	$0.000020 \pm$	0.000093	0.000019	± 0.000039	
canopy : dist_road	$0.000191 \pm$	0.000426	- 0.000183	± 0.000186 **	**
canopy : dist_forest_edge	$0.000209 \pm$	0.001433	0.000188	± 0.000488	
canopy : dist_forest_edge ²	- 0.000001 \pm	0.000011	- 0.000002	± 0.000004	



Figure 6. Predicted intensity of use for slope, during the day and during the night using the overall model. Intensity values were scaled between 0 and 1.



Figure 7. Predicted intensity of use for distance to the nearest road, during the day and during the night using the overall model. Intensity values were scaled between 0 and 1.

Figure 8. Predicted intensity of use for conifer proportion and distance to the nearest road as a function of canopy cover using the overall model. All other model variables were held at their mean. The intensity values were scaled between 0 and 1.

Figure 9. Percent of GPS locations which were located within the forest, using all locations (winter) or only a subset (weekend, workday, day, night). Weekend: Saturday and Sunday; workday: Tuesday and Thursday; day: 1h after sunrise to 1h before sunset; night: 2h after sunset to 2h before sunrise. Different letters indicate significant difference at P < 0.05.

We found only small differences between peri-urban and rural roe deer. While habitat selection of periurban and rural roe deer did nod differ during WE, WD and daytime (for all model variables: WE: 6 < W < 28, P > 0.073; WD: 7 < W < 25, P > 0.149; day: 8 < W < 27, P > 0.149), peri-urban roe deer avoided roads stronger during the night than rural roe deer did (*dist_road*: W = 32, P = 0.018; for all other model variables: 8 < W < 25, P > 0.149). Further, the difference of the odds ratios between WE and WD as well as between day and night did not differ between peri-urban and rural roe deer (for all model variables: WE & WD: 12 < W < 27, P > 0.149; day & night: 10 < W < 264, P > 0.343). This indicates that the magnitude of adaption between WE and WD as well as between day and night was similar for both groups of roe deer. Additionally, the percentage of GPS locations within the forest did not differ between peri-urban and rural roe deer in the whole winter (W = 18, P = 1), during the day (W = 19.5, P = 0.807) and during the night (W = 21, P = 0.639). Further, the significant higher use of open areas during nighttime than daytime did not differ between peri-urban and rural roe deer (W = 15, P = 0.755).

Discussion

Adaptation to the diel variations in human disturbance

Recreational activity in the peri-urban area of Zurich peaks during the day, declines towards twilight and is rare during the night (stationary visitor counting: 450 visitors per day; Wyttenbach 2012). In agreement with our predictions, the studied roe deer adapted HR and CA sizes and locations as well as habitat selection to the diel variation in human disturbance. The studied roe deer retreated to steeper habitats and avoided open areas more frequently during the day than during the night, likely as a result of the human disturbance during the day (slope: Imfeld 1996, Harris et al. 2014; open areas: Imfeld 1996, Náhlik et al. 2009, Bonnot et al. 2013). Consequently, roe deer reduced their moving range during the day which was also apparent in the HR and CA sizes. The diurnal HR and CA sizes were smaller than the nocturnal ones, likely as the roe deer intended to minimize the contact with human disturbance during the day (elk: Webb et al. 2011) and as a result of spending less time on open areas during the day compared to night (white-tailed deer [Odocoileus virginianus]: Walter et al. 2011, red deer: Richard et al. 2014). Beauchesne et al. (2014) found that mountain caribou expand their HR due to human disturbance until a disturbance threshold is reached. Above this threshold, individuals contract their HR while human disturbance still increases. Supposing roe deer exhibit a similar pattern, our findings would indicate that the recreational activity in the study area was on such a high level that the studied roe deer had to restrict their dispersal and had to retreat to suboptimal habitats (Beauchesne et al. 2014). The studied roe deer avoided areas up to 20 to 40 m from the roads which is in the range of some studies (moose 500 m: Laurian et al. 2008; red deer 20 to

50 m: Meisingset et al. 2013), but low compared to others (elk 1.8 km: Rowland et al. 2000; roe deer 2.5 km: Jiang et al. 2009). As the density of gravel roads and paths (mean \pm SD: 10.1 \pm 3.7 km/km²) as well as surfaced roads (1.8 \pm 1.6 km/km²) were very high within the HR of the studied roe deer, the roe deer could not shift to habitats which were further away from the roads, unless they would have left their HR. As the median value of distance to road within the HR was 16 m, the studied roe deer had limited access to more than 50% of their HR due to the avoidance of roads which might imply a loss of suitable habitat (Taylor et al. 2003, Fahrig et al. 2009) and thus could reduce population fitness (Nilsen et al. 2004).

General avoidance of roads

As recreational activity in the peri-urban area of Zurich mainly occurs during the day (Wyttenbach 2012), we assumed roe deer would avoid roads more strongly during the day than during the night. Further, we assumed that they would even select roads during the night, as road edges often offer high food supply (Klötzli 1965; but see Salek et al. 2010) which could attract ungulates (Meisingset et al. 2013). However, the studied roe deer avoided roads in general, independent of the time of day. In contrast to our results, Coulon et al. (2008) and Bonnot et al. (2013) found stronger avoidance of roads during the day than during the night on the level of the GPS locations. We suppose that the studied roe deer would have also exhibited stronger avoidance of roads during the day than during the day than higher sample size. Additionally, low human activity may have still occurred in the peri-urban area of Zurich during the night (stationary visitor counting: zero to one person per hour; Wyttenbach 2012) and disturbed the roe deer near roads. However, if such a disturbance occurred in the HR of the studied roe deer remains unclear and has to be further investigated. The studied roe deer not only avoided areas near (20 to 40 m) but also far away from roads (> 100 m). However, this is likely an artefact, because the areas with high distances to roads lay on the infrequent used border of the HR of only few studied roe deer.

Adaptation to the weekly and regional variations in human disturbance

Although the number of visitors in the peri-urban area of Zurich is about two times larger at WE than on WD (Wyttenbach 2012), the studied roe deer did neither exhibit differences regarding HR and CA sizes nor regarding habitat selection between WE and WD. Further, we analysed regional variation by comparing roe deer which were exposed to intense (peri-urban roe deer) with roe deer which were exposed to low recreational activity (rural roe deer). Peri-urban roe deer avoided roads more strongly than rural roe deer during the night. As the HR of the peri-urban roe deer comprised a higher density of surfaced roads than those of

the rural roe deer, it is likely that peri-urban roe deer were exposed to higher traffic and thus, experienced higher disturbance than the rural roe deer during the night. The lack of almost any weekly and regional difference may indicate the existence of a threshold value regarding the number of recreationists. We assume that below the threshold, the avoidance of roads, the retreating in steep slopes as well as the use of forest would have increased with increasing recreational activity and thus would have differed between WE and WD as well as between peri-urban and rural roe deer. Above a certain disturbance threshold, roe deer would have not changed their habitat selection, although the intensity of recreational activity increases. Colman et al. (2012) found such a disturbance threshold regarding the encounter rate between wild reindeer (*Rangifer tarandus tarandus*) and skiers. Below 133 skiers per day, the encounter rate between wild reindeer and skier increased linearly. However, wild reindeer avoided the ski trails and the area, within the radius of the upper quartile flight distance around them, completely if the number of skier exceed 133 per day. Thus, in our study area, recreational activity may have exceed a disturbance threshold during WE, WD and day as well as in the area of peri-urban and rural roe deer, but not during the night.

Tolerance to recreational activity

The studied roe deer adapted to the diel variation in recreational activity by modifying the use of steep slopes and open areas as well as of areas near roads with high canopy cover. Thus, our study confirmed findings of behavioural plasticity of roe deer regarding human disturbance (Hewison et al. 2001, Bonnot et al. 2013, Soennichsen et al. 2013). However, the difference between day and night also revealed that the studied roe deer were not tolerant to the human disturbance in our study area. We assume that roe deer which are tolerant to recreational activities would have used areas near roads as these areas usually offer high food supply (Klötzli 1965). Further, tolerant roe deer would have not retreated to steep slopes during the day. In contrast, the differences in the use of open areas between day and night may have been a result of the hunting activity in the study area (Borkowski et al. 2008). As hunting activities mainly occurred during daytime, roe deer may have sought concealment cover (Borkowski et al. 2008) and thus forest in this time period (Bonnot et al. 2013). Peri-urban roe deer may have tolerated higher human disturbance than rural roe deer which led to the lack of almost any differences between peri-urban and rural roe deer. Tolerance level increases over the course of a habituation process (Bejder et al. 2009). As habituation is a learning process over time, it cannot be investigated at one point in time (Bejder et al. 2009) and thus our study delivers no information about habituation. We can only assume that peri-urban may have stronger habituated than the rural roe deer to the recreational activity, as peri-urban roe deer may have perceived the

intense recreational activity in their HR more predictable than the rural roe deer did regarding the occasional activities in their HR (Reimers et al. 2011, Harris et al. 2014).

The effect of the habitat type

The habitat selection of roe deer did not only depend on roads or topography, but also on the habitat type. The studied roe deer generally selected for high conifer proportion within forest stands which is in line with the literature (Mysterud et al. 1995, Ratikainen et al. 2007). This selection pattern might be a consequence of enhanced concealment (Creel et al. 2005) and thermal cover (DelGiudice et al. 2013) under coniferous trees as well as of the forage preference for conifers during winter (Klötzli 1965, Cornelis et al. 1999). During the night, the studied roe deer selected high canopy cover, which offers thermal cover (Mysterud et al. 1995, Ratikainen et al. 2007, Ewald et al. 2014). However in winter, thermal cover may only occur in the combination with a high conifer proportion as broadleaved trees are bald. As we did not analyse habitat quality in our study area, we failed to disentangle the influence of habitat quality form those of recreational activity on the habitat selection of the studied roe deer. Tolerance and threshold values depend on habitat quality as the decision to shift from one habitat to another depends on the quality of the current and the alternative site (Gill et al. 2001, Harris et al. 2014). Animals inhabit habitats with high quality will less likely shift to habitats with poorer quality than animals would do which live already in low quality habitats. Further, animals which have high quality habitats as an alternative site will more likely shift to these sites than animals which have only low quality alternative sites (Harris et al. 2014). Future studies should account for habitat quality, for example by including LiDAR data in the analysis, as they deliver information about understory cover (Ewald et al. 2014) which may indicate concealment cover (Mysterud et al. 1999) and food supply (Ewald et al. 2014). Including habitat quality would be crucial for interpreting HR and CA shift, the avoidance of roads as well as habitat selection in general.

Conclusion and perspectives

Our study revealed that roe deer in a peri-urban landscape of Switzerland live in a landscape of fear shaped by non-motorized recreation activity as well as hunting. As a consequence, roe deer exhibit smaller CA and HR size during the day compared to the night, shift CA and HR between day and night, avoid roads in general as well as retreat in steeper areas and spend more time within the forest during the day than during the night. Thus, roe deer adapt to, but do not tolerate human disturbance. The lack of almost any difference between WE and WD as well as peri-urban and rural roe deer indicate the existence of a threshold value. However, as the responses of wildlife to human disturbance are very complex (Bejder et al. 2009), further research is needed to analyse this threshold values as well as to disentangle the influence of habitat quality and recreational activity on habitat selection of roe deer. Overall, our study indicates that roe deer response to recreational activity and may have limited access to a considerable amount of habitats due to the disturbance. This may lead to a negative energy budget (Houston et al. 2012) and thus presumably has an effect on the individual and the population fitness (Ciuti et al. 2012). Thus, it is essential that future studies also investigate the consequences of the observed modifications in habitat selection on the fitness of individual roe deer as well as the entire population.

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Appendix I: home range and core area data as well as additional analyses

Table 1. Individual home range and core area sizes using all GPS locations (winter), locations from Saturday and Sunday (weekend) or locations from Tuesday and Thursday (workday). Sample [#] contains the number of GPS location used for the particular analysis. The column h [m] contains the value of the smoothing parameter $h_{ad \ hoc}$. Statistics were conducted among roe deer which were exposed to intense (peri-urban roe deer) or low recreational activities (rural roe deer) and among all roe deer (all).

		v	/inter			Wee	ekend		Workday			
Animal ID	CA [ha]	HR [ha]	Sample [#]	h [m]	CA [ha]	HR [ha]	Sample [#]	h [m]	CA [ha]	HR [ha]	Sample [#]	h [m]
RD01	7.11	30.54	1400	46	7.10	29.55	402	50	7.37	30.26	398	50
RD02	21.79	98.96	1423	98	17.84	81.82	403	96	24.03	109.47	408	111
RD03	4.02	18.20	1432	31	4.02	17.59	408	33	4.19	17.87	411	34
RD04	4.68	25.72	1411	41	6.07	33.36	397	49	4.43	19.19	403	40
RD05	3.59	14.41	1434	28	3.55	14.16	405	32	3.64	14.21	409	31
RD06	3.33	13.29	1223	27	3.36	13.44	332	30	3.53	13.83	363	30
RD07	5.38	54.75	1402	58	5.15	48.10	398	57	6.82	58.17	398	67
RD08	7.72	29.01	1404	44	8.72	30.75	404	49	7.14	29.04	401	48
RD09	9.35	51.79	1238	58	12.59	71.58	356	78	8.67	47.10	358	57
RD10	10.54	44.45	1192	54	12.97	51.20	343	63	9.62	40.62	343	57
RD11	19.26	107.61	1173	81	21.85	111.98	330	95	17.51	98.21	339	82
RD12	8.46	45.64	1112	59	9.13	44.67	315	65	8.31	41.84	320	65
mean: peri-urban	10.66	54.87	1306	63	10.44	51.07	372	66	11.23	56.07	373	70
mean: rural	7.42	37.15	1331	44	8.59	41.84	376	52	7.02	34.21	383	46
mean: all	8.77	44.53	1320	52	9.36	45.68	374	58	8.77	43.32	379	56
SD: peri-urban	6.50	26.12	144	20	5.05	19.10	41	18	7.23	31.48	39	24
SD: rural	5.69	33.72	114	19	6.73	36.88	35	25	5.02	30.54	29	19
SD: all	5.99	30.85	121	21	5.91	29.95	36	23	6.12	31.56	33	23

Table 2. Individual HR and CA sizes using either diurnal, nocturnal or twilight GPS locations. Day: 1h after sunrise to 1h before sunset; night: 2h after sunset to 2h before sunrise; twilight: between day and night. Sample [#] contains the number of GPS location used for the particular analysis. The column h [m] contains the value of the smoothing parameter $h_{ad hoc}$. Statistics were conducted among roe deer which were exposed to intense (peri-urban roe deer) or low recreational activities (rural roe deer) and among all roe deer (all).

		C	Day			N	ight		Twilight				
Animal ID	CA [ha]	HR [ha]	Sample [#]	h [m]	CA [ha]	HR [ha]	Sample [#]	h [m]	CA [ha]	HR [ha]	Sample [#]	h [m]	
RD01	4.81	27.18	324	46	8.98	31.20	374	50	7.63	31.91	352	52	
RD02	21.56	94.19	330	107	21.75	104.26	380	101	24.99	107.56	358	112	
RD03	1.31	7.08	336	21	6.85	22.80	378	39	4.76	19.37	361	36	
RD04	5.28	30.90	330	52	3.65	15.96	374	33	5.23	30.96	355	49	
RD05	0.95	6.34	335	21	3.65	15.74	381	31	4.50	17.15	359	34	
RD06	0.72	5.08	299	18	5.01	16.09	313	34	4.21	13.53	304	31	
RD07	3.70	47.17	325	60	6.74	57.14	375	60	5.78	57.90	351	66	
RD08	3.04	14.88	326	31	4.83	26.27	375	39	8.50	29.52	352	48	
RD09	6.59	38.81	283	67	7.90	47.78	333	54	9.58	55.81	313	64	
RD10	6.79	31.06	270	54	7.77	38.51	324	54	8.79	38.84	300	55	
RD11	11.48	74.98	269	77	17.94	112.92	318	89	19.58	110.02	292	90	
RD12	4.17	33.00	254	57	9.06	38.59	303	61	8.77	39.07	276	64	
mean: peri-urban	8.21	46.52	301	65	10.86	53.94	351	65	11.19	55.05	327	70	
mean: rural	4.20	25.44	311	41	7.12	36.80	353	46	8.05	39.48	334	50	
mean: all	5.87	34.22	307	51	8.68	43.94	352	54	9.36	45.97	331	58	
SD: peri-urban	7.56	27.70	36	24	6.16	29.72	35	20	7.81	30.90	37	24	
SD: rural	3.92	25.47	27	24	5.03	35.42	30	21	5.50	34.13	30	21	
SD: all	5.78	27.40	30	26	5.60	32.92	31	22	6.43	32.36	31	24	

Table 3. Mean percent overlap [%] between weekend and workday (WE/WD), day and night, day and twilight as well as night and twilight using either individual HR or CA. Statistics were conducted among roe deer which were exposed to intense recreational activities (peri-urban roe deer), among roe deer which were exposed to low recreational activities (rural roe deer) and among all roe deer (all).

		Mean overl	ap of HR [%]			Mean overla	p of CA [%]	
Animal ID	WE/WD	Day/night	Day/twilight	Night/twilight	WE/WD	Day/night	Day/twilight	Night/twilight
RD01	95	84	87	94	70	73	82	76
RD02	86	84	92	90	84	63	81	79
RD03	94	66	68	92	91	37	64	50
RD04	79	75	83	76	84	63	77	81
RD05	95	69	68	92	80	0	56	57
RD06	90	66	69	91	77	1	19	64
RD07	84	77	82	88	88	53	80	72
RD08	93	55	74	87	81	0	68	32
RD09	81	77	82	85	81	31	57	74
RD10	89	56	81	74	81	5	62	33
RD11	80	66	80	82	88	21	67	56
RD12	80	62	68	87	79	28	74	49
mean: peri-urban	87	72	82	87	80	44	76	62
mean: rural	87	68	75	86	83	22	58	59
mean: all	87	70	78	87	82	31	65	60
SD: peri-urban	6	13	9	8	7	28	8	20
SD: rural	7	7	7	6	5	24	19	16
SD: all	6	10	8	6	6	27	17	17

Variation in the core area overlap day/night

Core area overlap between day and night had a high variance which we could neither explain with the classification peri-urban and rural roe deer (F-statistics = 2.30, P = 0.161), nor with the sex (F-statistics = 0.09, P = 0.766), the density of surfaced roads (F-statistics = 0.39, P = 0.546), the density of gravel roads and paths (F-statistics = 0.06, P = 0.811), the percentage of GPS locations within the forest (winter: F-statistics = 0.13, P = 0.725; day: F-statistics = 2.53, P = 0.143; night: F-statistics = 0.01, P = 0.937), the difference between the diurnal and nocturnal percentage of GPS locations within the forest (F-statistics = 0.09, P = 0.775) or the mean of any habitat variable in the HR (*canopy*: F-statistics = 0.39, P = 0.545; *conifers*: F-statistics = 0.01, P = 0.933; *slope*: F-statistics = 0.22, P = 0.649; *dist_road*: F-statistics = 1.29, P = 0.282; *dist_forest_edge*: F-statistics = 3.66, P = 0.085). Further also the difference between the odds ratios of day and night could not explain this variance (for all model variables: 0.00 < F-statistics < 1.83, P > 0.206).

Appendix II: characterisation of the home ranges

Table 4. Mean of the habitat variables within the individual HR. Canopy, conifer, slope, dist_road and dist_forest_edge are variables used in the resource selection function. Canopy: canopy cover of tree crowns [%]; conifers: conifer proportion [%]; slope [°]; dist_road: distance to the nearest road [m]; dist_build: distance to the nearest building [m], dist_forest_edge: distance to the nearest forest edge [m]. Further, the percentage of forested areas as well as the density of surfaced and gravel roads is shown. Gravel roads include gravel paths many of which are closed to public traffic. Statistics were conducted among roe deer which were exposed to intense recreational activities (peri-urban roe deer), among roe deer which were exposed to low recreational activities (rural roe deer) and among all roe deer (all).

Animal ID	Canopy [%]	Conifer [%]	Slope [°]	Dist_road [m]	Dist_forest_	Forest	Surfaced roads	Gravel roads
					edge [m]	proportion [%]	[m/ha]	[m/ha]
RD01	63	58	8	30	69	88	30	77
RD02	42	15	14	27	60	55	26	130
RD03	45	11	14	49	39	58	0	78
RD04	58	11	19	31	59	70	0	89
RD05	69	47	4	31	144	94	11	154
RD06	66	43	4	30	120	92	34	153
RD07	55	18	9	26	31	67	48	144
RD08	47	6	21	73	35	61	2	62
RD09	32	3	17	73	54	36	0	41
RD10	56	23	14	43	96	66	18	72
RD11	58	12	15	34	110	70	12	94
RD12	73	49	8	20	59	89	38	125
mean: peri-urban	58	33	11	29	63	73	32	110
mean: rural	47	20	16	57	75	60	20	96
mean: all	55	25	12	39	73	71	18	101
SD: peri-urban	10	17	3	8	21	13	10	29
SD: rural	19	17	10	28	38	27	32	43
SD: all	11	18	5	17	35	17	16	37

The HR of the peri-urban and the rural roe deer did not differ regarding the mean of canopy cover (W = 19, P = 0.876), conifer proportion (W = 29, P = 0.073), slope (W = 10, P = 0.268), distance to the nearest forest edge (W = 15, P = 0.755), forest proportion within the HR (W = 17, P = 1) and the density of gravel roads (W = 19, P = 0.876). In contrast, the HR of the peri-urban roe deer had a higher density of surfaced roads (W = 32, P = 0.022) and lower distances to the nearest road (W = 5, P = 0.048) than the HR of the rural roe deer.

Appendix III: data and additional analyses of the habitat selection models

Table 5, 6, 7, 8 (page 38 to 41). Results of the individual resource selection functions using GPS locations from either weekend, workday, day or night. The model quality is illustrated by the theta, the dispersion and the McFadden's pseudo R². The diurnal and nocturnal models of RD10 and RD12 were calculated using the VGAM package (Thomas et al. 1996) in the software R 3.1.1 (R Core Team 2014). We could not extract dispersion and McFadden's pseudo R² for this models which is indicated with a star (*). The sample size for the models equaled the number of circles within the home range of the individual roe deer. Further, the tables show the percentage of circles per roe deer which contain zero GPS locations as well as the mean number of GPS locations within a circle. As not all GPS locations are located within a circle unit, the percentage of GPS locations per roe deer that are located within a circle is given.

Table 5. Table caption see page 37.

		Weekend											
		RD01	RD02	RD03	RD04	RD05	RD06	RD07	RD08	RD09	RD10	RD11	RD12
Intercept and coefficient	<pre>(Intercept) canopy [%] conifers [%] conifers² slope [°] dist_road [m] dist_road² dist_forest_edge [m] dist_forest_edge² canopy : conifers canopy : conifers² canopy : dist_road canopy : dist_forest_edge²</pre>	-8.119994 0.006307 -0.001903 0.003154 0.035762 -0.000267 0.015971 -0.000113 0.000155 -0.00008 -0.00008 -0.00008 0.000008	-10.713220 0.027666 0.245740 -0.003887 0.034034 0.075943 -0.000306 -0.0003931 -0.000279 -0.002822 0.000046 -0.000355 0.000354 -0.000002	-7.617670 0.008450 0.302216 -0.007930 -0.006105 0.041389 -0.000182 0.006381 -0.000180 -0.003751 0.000092 -0.000019 -0.000174 0.000003	-6.461957 -0.000519 0.119005 -0.000147 -0.004499 0.035861 -0.000270 -0.074374 0.000436 -0.001069 -0.000002 0.000072 0.000072 -0.000075	-9.681164 0.042385 0.067857 -0.000424 -0.067461 0.075108 -0.000528 0.043047 -0.000196 -0.001007 0.000007 -0.000344 -0.000442 0.000002	-8.753712 0.039031 -0.016485 0.000398 -0.214244 0.025198 0.000017 0.060380 -0.000234 0.000184 -0.00005 -0.000208 -0.000294 0.000002	-7.705725 -0.010438 0.230458 -0.003847 -0.115152 0.097346 -0.000760 -0.059864 0.000408 -0.003238 0.000049 0.000362 0.001167 -0.000014	-5.645945 -0.004678 0.113627 -0.001606 -0.047626 0.009234 -0.000037 -0.088100 0.000812 -0.001397 0.000021 0.000057 0.001442 -0.000013	-11.138297 0.026286 -0.403547 0.047260 0.090609 0.030348 -0.000079 0.019049 -0.000076 0.005631 -0.000570 -0.000292 -0.000404 0.000002	-7.608283 -0.003320 0.074149 0.00099 -0.012836 0.028855 -0.000150 0.002301 -0.00036 -0.000312 -0.00008 -0.00001 0.000129 -0.000001	-9.473768 0.012033 0.062394 -0.000657 -0.034456 0.013561 -0.00003 0.037355 -0.000194 -0.0000554 0.000005 -0.000255 0.000025	-16.266001 0.080138 -0.031913 0.000308 0.050649 0.142124 -0.002099 0.154490 -0.000531 0.000190 -0.000022 -0.000024 -0.000374 -0.001268 0.000003
SE of Intercept and coefficients	<pre>(Intercept) canopy [%] conifers [%] conifers² slope [°] dist_road [m] dist_road² dist_forest_edge [m] dist_forest_edge² canopy : conifers canopy : conifers² canopy : dist_road canopy : dist_forest_edge²</pre>	0.470795 0.006914 0.036253 0.000376 0.016645 0.016211 0.000154 0.000120 0.000532 0.000006 0.000125 0.000291 0.000002	0.885788 0.011126 0.076543 0.001064 0.021083 0.000179 0.029920 0.000300 0.001000 0.00013 0.000163 0.000421 0.00005	0.460867 0.005934 0.058844 0.002039 0.019590 0.009366 0.000065 0.011970 0.000116 0.000091 0.000023 0.000023 0.0000223 0.000022	0.596325 0.006716 0.057619 0.01353 0.019923 0.000224 0.020980 0.000201 0.000713 0.000016 0.000122 0.000279 0.000003	1.119604 0.014400 0.044472 0.000494 0.066286 0.022993 0.000201 0.018142 0.000065 0.0000593 0.000006 0.000210 0.000239 0.000001	1.010901 0.012923 0.038064 0.000439 0.063930 0.019284 0.000170 0.018275 0.000074 0.000510 0.00006 0.000178 0.000252 0.000001	0.510763 0.008140 0.090879 0.001793 0.017013 0.019702 0.000191 0.024459 0.000260 0.001133 0.000021 0.000138 0.000432 0.000005	0.458702 0.006201 0.078184 0.001325 0.020818 0.005527 0.000020 0.021671 0.000255 0.001299 0.000035 0.0000035 0.0000003 0.000003	0.550167 0.010494 0.309212 0.025089 0.017333 0.007875 0.000038 0.007343 0.000035 0.004753 0.000344 0.000045 0.000264 0.000003	0.457017 0.007323 0.045357 0.000771 0.012033 0.006561 0.000034 0.007473 0.000031 0.000605 0.000010 0.000059 0.000135 0.000001	0.515893 0.006577 0.083584 0.001041 0.012387 0.008938 0.000072 0.010666 0.000058 0.000916 0.000012 0.000072 0.000132 0.0000132	3.130383 0.036117 0.092093 0.000881 0.017457 0.063159 0.000525 0.080902 0.000542 0.001090 0.000010 0.000010 0.000050 0.000950 0.000006
model quality	theta dispersion MCFadden's pseudo R²	1.16 0.86 0.19	0.41 0.36 0.45	2.45 1.03 0.30	0.75 0.82 0.22	1.65 1.06 0.29	4.73 1.11 0.30	0.29 0.43 0.39	1.03 0.85 0.17	0.65 0.53 0.29	0.73 0.64 0.19	0.60 0.42 0.18	0.71 0.57 0.29
statistics	sample size (number of circles) circles with zero location [%] mean locations per circle locations within circles [%]	478.00 63.39 0.67 81.17	1103.00 86.85 0.28 77.19	289.00 50.52 1.04 75.44	418.00 65.55 0.74 81.70	228.00 46.05 1.34 77.47	210.00 42.38 1.19 77.33	763.00 85.19 0.39 75.76	460.00 64.13 0.66 76.32	809.00 81.71 0.31 75.53	702.00 77.92 0.36 76.19	1634.00 88.25 0.16 81.88	672.00 79.46 0.35 76.64

Table 6. Table caption see page 37.

			Workday										
		RD01	RD02	RD03	RD04	RD05	RD06	RD07	RD08	RD09	RD10	RD11	RD12
Intercept and coefficients	<pre>(Intercept) canopy [%] conifers [%] conifers² slope [°] dist_road [m] dist_road² dist_forest_edge [m] dist_forest_edge² canopy : conifers canopy : conifers² canopy : dist_road canopy : dist_forest_edge²</pre>	-9.526732 0.010095 0.110822 -0.001101 0.088639 0.055053 -0.000297 0.000774 -0.000039 -0.001238 0.000012 -0.000219 0.000348 -0.000002	-11.045208 0.024987 0.155510 -0.001948 0.041508 0.106077 -0.000652 -0.017456 -0.000245 -0.001440 0.000020 -0.000348 0.000032 0.000003	-8.306699 0.013111 0.417879 -0.011381 0.005044 0.045317 -0.000164 0.003883 -0.000160 -0.004928 0.000130 -0.000123 0.000140 -0.000001	-6.137533 0.001981 0.139273 -0.045090 0.080229 -0.000775 -0.119187 0.000862 -0.001369 0.000003 -0.000179 0.001251 -0.000010	-9.427142 0.029358 -0.035125 0.000771 -0.060109 0.073887 -0.000510 0.058605 -0.000234 0.000579 -0.000011 -0.000361 -0.000573 0.000023	-8.048056 0.021809 -0.035214 0.000778 -0.124260 0.019730 -0.000148 0.044923 -0.000170 0.000512 -0.000010 -0.000011 -0.000375 0.000001	-7.403540 0.001506 0.175858 -0.001738 -0.090929 0.077247 -0.000557 -0.041085 0.000156 -0.002503 0.000025 0.000083 0.0000353 -0.000004	-5.322612 -0.009080 0.046683 -0.001077 -0.86200 0.015548 -0.000051 -0.063531 0.000454 -0.000864 0.000035 0.000004 0.001509 -0.000011	-10.662380 0.012198 -0.322054 0.023217 0.114747 0.023168 -0.000054 0.008232 -0.000025 0.004373 -0.000274 -0.000237 0.000102 -0.000001	-7.594949 0.016301 0.131308 -0.001501 -0.040423 0.032503 -0.000177 0.002801 -0.00036 -0.001423 0.000016 -0.000039 -0.000053 0.000000	-9.079698 -0.002144 0.048702 0.000019 -0.005804 0.020564 0.000022 0.007387 -0.000036 -0.000185 -0.00001 -0.000043 -0.000006 0.000000	-10.913895 0.017267 -0.028508 0.000213 0.106115 0.008126 -0.000243 0.088744 -0.000314 0.000361 -0.000033 0.000115 -0.000577 0.000002
SE of Intercept and coefficients	<pre>(Intercept) canopy [%] conifers [%] conifers² slope [°] dist_road [m] dist_road² dist_forest_edge [m] dist_forest_edge² canopy : conifers canopy : dist_road canopy : dist_forest_edge canopy : dist_forest_edge²</pre>	0.675195 0.009168 0.0040097 0.017652 0.018764 0.000166 0.023833 0.000143 0.000567 0.000066 0.000151 0.000329 0.000002	0.809206 0.009592 0.064919 0.000893 0.012463 0.019276 0.000157 0.022064 0.000230 0.000842 0.000011 0.000150 0.000274 0.000003	0.493402 0.006144 0.059936 0.002026 0.019286 0.009662 0.012315 0.000119 0.000696 0.000023 0.000023 0.000057 0.000028 0.000002	0.689478 0.007840 0.067583 0.001608 0.018056 0.024492 0.000284 0.0025192 0.000242 0.000842 0.000019 0.000147 0.0000337 0.000003	0.973321 0.012531 0.039603 0.000451 0.064979 0.020584 0.000184 0.015498 0.000056 0.00056 0.0000542 0.00006 0.0000181 0.000206 0.000001	0.900969 0.011865 0.039054 0.000443 0.059348 0.019389 0.00180 0.018160 0.000074 0.000528 0.00001 0.000187 0.000257 0.000001	0.481717 0.007566 0.072766 0.015976 0.018825 0.000185 0.025115 0.000295 0.000953 0.000953 0.000011 0.000133 0.000426 0.000006	0.443642 0.005870 0.071899 0.001134 0.021122 0.005264 0.000020 0.019627 0.000233 0.001092 0.000028 0.000057 0.000278 0.00003	0.478866 0.006661 0.253593 0.013164 0.016560 0.006907 0.000034 0.000034 0.000025 0.002898 0.000146 0.000040 0.000351 0.000006	0.494009 0.007174 0.048321 0.00781 0.012464 0.007374 0.000040 0.007745 0.000031 0.000622 0.000010 0.000065 0.000125 0.000001	0.439163 0.005969 0.080658 0.000965 0.012699 0.008867 0.000071 0.008373 0.000043 0.000844 0.000011 0.00067 0.000101 0.000000	1.149187 0.014113 0.080409 0.000818 0.017898 0.038147 0.000335 0.046690 0.000290 0.000962 0.000962 0.000010 0.000343 0.000538 0.000003
model quality	theta dispersion MCFadden's pseudo R²	0.97 0.82 0.20	0.74 0.43 0.43	2.47 1.04 0.35	0.54 0.72 0.28	2.55 1.10 0.29	3.18 1.12 0.30	0.26 0.46 0.29	1.17 0.86 0.21	0.78 0.57 0.27	0.58 0.64 0.14	0.64 0.42 0.22	0.53 0.58 0.24
statistics	sample size (number of circles) circles with zero location [%] mean locations per circle locations within circles [%]	478.00 66.53 0.63 78.29	1103.00 84.86 0.27 74.94	289.00 51.90 1.11 80.30	418.00 70.57 0.72 75.69	228.00 42.11 1.36 77.69	210.00 42.38 1.30 77.34	763.00 84.27 0.40 76.77	460.00 64.78 0.65 76.40	809.00 79.85 0.33 75.71	702.00 77.35 0.37 77.06	1634.00 88.37 0.17 81.82	672.00 79.17 0.38 80.89

Table 7. Table caption see page 37.

							Day	,					
		RD01	RD02	RD03	RD04	RD05	RD06	RD07	RD08	RD09	RD10	RD11	RD12
	(Intercept)	-9.664377	-11.625965	-12.332791	-8.390393	-10.096186	-9.907443	-6.953584	-9.369443	-9.858073	-11.205919	-13.988765	-25.271188
	canopy [%]	0.011702	0.042745	0.023806	0.003895	-0.003065	-0.006872	-0.029872	-0.001039	-0.020067	0.034759	0.031617	0.161100
	conifers [%]	0.041706	0.427216	0.774254	0.307712	0.025160	0.004818	0.416185	0.128388	0.021177	0.304589	0.384879	0.426458
nts	conifers ²	-0.000353	-0.005074	-0.017960	-0.003237	0.000609	0.000859	-0.004019	-0.001872	0.010940	-0.003770	-0.003219	-0.003590
icie	slope [°]	0.115073	0.059469	-0.009693	0.010387	0.023199	-0.175607	-0.071011	-0.026537	0.153983	0.022531	0.040217	0.106096
beff	dist_road [m]	0.068678	0.068181	0.075395	0.130967	0.059958	0.028420	0.136703	0.038487	0.016676	0.016172	0.064996	-0.026991
d d	dist_road ²	-0.000649	-0.000231	-0.000303	-0.001613	-0.000704	-0.000286	-0.001375	-0.000107	-0.000092	-0.000129	-0.000115	-0.000627
an	dist_forest_edge [m]	0.032465	-0.077034	0.139857	-0.126840	0.086706	0.085278	-0.191594	-0.092798	-0.154378	0.037662	-0.009623	0.299650
ept	dist_forest_edge ²	-0.000233	-0.000045	-0.001943	0.000771	-0.000381	-0.000407	0.001119	0.000868	0.000457	-0.000463	0.000023	-0.001518
ero	canopy : conifers	-0.000294	-0.004958	-0.009438	-0.003511	0.000088	0.000452	-0.005193	-0.001089	0.003172	-0.002799	-0.003891	-0.005486
Int	canopy : conifers ²	0.000000	0.000058	0.000204	0.000035	-0.000010	-0.000014	0.000048	0.000023	-0.000214	0.000035	0.000032	0.000047
	canopy : dist_road	0.000074	-0.000373	0.000018	-0.000003	0.000450	0.000484	0.000629	0.000034	-0.000066	0.000098	-0.000228	0.001178
	canopy : dist_forest_edge	-0.000212	0.000675	-0.000713	0.001420	-0.000929	-0.000695	0.002232	0.002005	0.001499	-0.000363	0.000278	-0.002691
	canopy : dist_forest_edge ²	0.000001	0.000001	0.000015	-0.000009	0.000004	0.000004	-0.000019	-0.000018	-0.000005	0.000004	-0.000001	0.000014
	(Intercept)	0.641756	1.335984	0.919643	0.955622	1.250595	1.525987	0.762376	0.820016	0.646709	1.467386	1.475512	6.003997
	canopy [%]	0.009047	0.015926	0.011599	0.010776	0.018053	0.022047	0.011857	0.010246	0.008607	0.018093	0.017290	0.070348
Its	conifers [%]	0.044138	0.095786	0.092229	0.082941	0.056172	0.060982	0.110325	0.089717	0.254802	0.080616	0.099974	0.152365
cier	conifers ²	0.000457	0.001282	0.002946	0.001972	0.000645	0.000692	0.001307	0.001313	0.011968	0.001288	0.001192	0.001300
effi	slope [°]	0.020277	0.015183	0.027000	0.021283	0.116327	0.123132	0.019818	0.026721	0.017957	0.015154	0.012492	0.023315
Ō	dist_road [m]	0.020851	0.028479	0.014655	0.031646	0.025371	0.027620	0.027294	0.009067	0.008783	0.013023	0.017579	0.060767
bue	dist_road ²	0.000202	0.000210	0.000093	0.000382	0.000254	0.000262	0.000283	0.000024	0.000042	0.000041	0.000078	0.000414
pt	dist_forest_edge [m]	0.027633	0.041916	0.029763	0.036552	0.023123	0.032357	0.066641	0.040126	0.035622	0.042910	0.018617	0.095749
irce	dist_forest_edge ²	0.000162	0.000342	0.000386	0.000308	0.000088	0.000137	0.001264	0.000350	0.000160	0.000306	0.000081	0.000534
nte	canopy : conifers	0.000627	0.001230	0.001236	0.001014	0.000830	0.000862	0.001496	0.001344	0.003321	0.000975	0.001121	0.001811
of I	canopy : conifers ²	0.000007	0.000016	0.000037	0.000023	0.000009	0.000010	0.000017	0.000032	0.000154	0.000015	0.000014	0.000016
SE	canopy : dist_road	0.000162	0.000251	0.000092	0.000207	0.000278	0.000310	0.000196	0.000091	0.000056	0.000130	0.000171	0.000674
	canopy : dist_forest_edge	0.000374	0.000513	0.000431	0.000469	0.000321	0.000479	0.000918	0.000532	0.000440	0.000503	0.000216	0.001096
	canopy : dist_forest_edge ²	0.000002	0.000004	0.000005	0.000004	0.000001	0.000002	0.000016	0.000005	0.000003	0.000003	0.000001	0.000006
le Vi	theta	0.53	0.27	1.25	0.33	1.38	1.36	0.12	0.77	0.87	0.00	0.46	0.00
ode iali	dispersion	0.73	0.31	0.62	0.60	0.77	0.79	0.29	0.57	0.38	*	0.31	*
a n	MCFadden's pseudo R ²	0.27	0.51	0.72	0.37	0.69	0.68	0.45	0.56	0.64	*	0.46	*
10	sample size (number of circles)	478.00	1103.00	289.00	418.00	228.00	210.00	763.00	460.00	809.00	702.00	1634.00	672.00
stics	circles with zero location [%]	69.46	87.85	69.90	74.40	65.35	65.71	89.52	75.00	84.92	83.19	89.41	84.97
atis	mean locations per circle	0.78	0.33	1.42	0.90	1.65	1.59	0.52	0.77	0.38	0.43	0.20	0.44
<u> </u>	•										75.00		~~~~

Table 8. Table caption see page 37.

							Nigh	t					
		RD01	RD02	RD03	RD04	RD05	RD06	RD07	RD08	RD09	RD10	RD11	RD12
Intercept and coefficients	<pre>(Intercept) canopy [%] conifers [%] conifers² slope [°] dist_road [m] dist_road² dist_forest_edge [m] dist_forest_edge² canopy : conifers canopy : conifers² canopy : dist_road canopy : dist_forest_edge²</pre>	-8.040834 0.008437 0.133618 -0.001222 0.048620 0.033860 -0.000124 -0.012870 0.000080 -0.001845 0.000017 -0.000257 0.000257 0.000410 -0.00003	-9.789945 0.013714 0.064061 -0.001050 0.025706 0.084509 -0.000577 -0.020612 -0.000009 -0.000822 0.000015 -0.000066 0.000253 -0.000001	-6.858275 0.003715 0.097354 -0.003297 0.020204 0.028747 -0.000124 -0.018258 0.000080 -0.000763 0.000033 -0.000127 -0.000085 0.000001	-5.128982 -0.004071 0.046846 0.001094 -0.036926 0.033319 -0.000244 -0.102556 0.000617 -0.000030 -0.000018 -0.000233 0.000968 -0.00006	-8.432230 0.026189 -0.016480 0.000084 0.006584 0.044679 -0.000405 0.031387 -0.000122 0.000212 -0.000001 -0.000192 -0.000264 0.000001	-6.590753 0.020773 0.014241 -0.000234 -0.121729 0.015750 -0.000316 0.011050 -0.000037 -0.000339 0.000004 0.000087 -0.000121 0.000000	-8.322029 0.008058 0.144126 -0.001802 -0.113558 0.071243 -0.000370 -0.005512 0.000000 -0.002180 0.000026 0.000021 0.000021 0.000031 0.000416 -0.000007	-4.209945 -0.000644 0.057150 -0.000376 -0.119812 0.013639 -0.000061 -0.061021 0.000353 -0.000865 0.000009 -0.000054 0.001099 -0.000007	-10.016509 0.038121 0.384628 -0.012943 0.043696 0.038580 -0.000117 0.007886 -0.000023 -0.003275 0.000132 -0.000512 0.0000512 0.000002	-7.037813 0.012396 0.062868 0.00030 -0.093124 0.041572 -0.000240 0.002093 -0.000020 -0.000354 -0.000055 -0.000163 -0.000026 -0.000001	-8.035045 0.003180 0.169990 -0.001988 -0.081199 0.002801 0.002801 0.0023634 -0.000117 -0.001580 0.000022 -0.000196 0.000096 0.000001	-9.230240 0.000162 -0.129830 0.000706 0.072874 0.102109 -0.001609 0.011225 0.000374 0.001701 -0.00010 -0.000481 0.000311 -0.00007
SE of Intercept and coefficients	<pre>(Intercept) canopy [%] conifers [%] conifers² slope [°] dist_road [m] dist_road² dist_forest_edge [m] dist_forest_edge² canopy : conifers canopy : conifers² canopy : dist_road canopy : dist_forest_edge²</pre>	0.465921 0.007018 0.037280 0.016975 0.015940 0.000150 0.020506 0.000127 0.000555 0.00006 0.000125 0.00006 0.000125 0.000294 0.000002	0.579566 0.007269 0.069117 0.000945 0.012925 0.017341 0.000164 0.016667 0.000138 0.000900 0.000012 0.000128 0.000226 0.000002	0.392117 0.005071 0.060102 0.002079 0.018880 0.008506 0.000061 0.0009604 0.000084 0.000084 0.000082 0.000023 0.000023 0.0000195 0.000002	0.572304 0.006634 0.064582 0.001500 0.016317 0.019815 0.000227 0.020668 0.000209 0.000786 0.00018 0.00018 0.000124 0.000281 0.00003	0.949861 0.012310 0.048253 0.063763 0.022614 0.000211 0.017230 0.000061 0.000061 0.000025 0.000007 0.000227 0.000021	0.701616 0.009403 0.040112 0.000457 0.051187 0.020019 0.000199 0.016060 0.000064 0.0000524 0.00006 0.000186 0.000227 0.000001	0.464758 0.007002 0.064936 0.000836 0.014955 0.016966 0.000160 0.020354 0.000216 0.000832 0.000010 0.000119 0.000358 0.000004	0.470823 0.006876 0.092903 0.001647 0.023231 0.005715 0.000026 0.021152 0.000328 0.001585 0.000045 0.000045 0.000061 0.000312 0.000004	0.466675 0.006776 0.298832 0.013574 0.017109 0.007526 0.000040 0.005393 0.000022 0.003358 0.000149 0.000056 0.000314 0.000005	0.451536 0.007172 0.055788 0.000867 0.015091 0.008707 0.000059 0.006163 0.000022 0.000736 0.000011 0.000078 0.000122 0.00001	0.373723 0.005248 0.080596 0.001061 0.015061 0.008658 0.00078 0.000738 0.00039 0.00039 0.000901 0.000091 0.000064 0.000097 0.000000	0.665379 0.008477 0.073897 0.000786 0.035210 0.000435 0.038960 0.000312 0.000868 0.0000276 0.000276 0.000468 0.00004
model quality	theta dispersion MCFadden's pseudo R²	0.72 0.89 0.13	0.35 0.46 0.35	1.81 1.13 0.14	0.52 0.81 0.25	0.94 1.10 0.12	1.62 1.17 0.12	0.35 0.55 0.32	0.60 0.81 0.24	0.46 0.58 0.29	0.00 * *	0.38 0.40 0.29	0.00 * *
statistics	sample size (number of circles) circles with zero location [%] mean locations per circle locations within circles [%]	478.00 57.95 0.92 81.00	1103.00 82.86 0.40 79.17	289.00 38.41 1.43 76.58	418.00 64.83 1.04 77.78	228.00 38.16 1.89 77.98	210.00 36.19 1.64 76.50	763.00 80.08 0.53 73.33	460.00 61.96 0.95 79.13	809.00 78.12 0.47 77.14	702.00 76.21 0.55 80.38	1634.00 87.58 0.22 78.48	672.00 74.26 0.52 78.83

Table 9. Differences between the odds ratios of the weekend and the workday models. The difference was analysed using Wilcoxon rank sum test. A p-value < 0.05 indicates significance difference between WE and WD models concerning the particular model variable.

Model variable	RD01	RD02	RD03	RD04	RD05	RD06	RD07	RD08	RD09	RD10	RD11	RD12	v v	P-value
canopy [%]	0.003820	-0.002750	0.004711	0.002502	-0.013502	-0.017754	0.011891	-0.004372	-0.014362	0.019749	-0.014247	-0.066019	26	0.339
conifers [%]	0.119097	-0.110314	0.165884	0.023062	-0.104728	-0.018251	-0.066908	-0.072544	0.056713	0.063351	-0.014474	0.003304	39	1.000
conifers ²	-0.001439	0.001933	-0.003418	-0.000364	0.001196	0.000381	0.002102	0.000529	-0.024905	-0.001599	0.000676	-0.000095	37	0.910
slope [°]	0.027495	0.007762	0.011143	-0.039600	0.006898	0.075999	0.021851	-0.036080	0.026748	-0.026863	0.028082	0.059997	53	0.301
dist_road [m]	0.020188	0.033007	0.004101	0.047024	-0.001316	-0.005593	-0.021933	0.006393	-0.007375	0.003762	0.007123	-0.144560	45	0.677
dist_road ²	-0.000030	-0.000346	0.000018	-0.000504	0.000018	-0.000165	0.000202	-0.000014	0.000025	-0.000027	0.000025	0.001854	35	0.791
dist_forest_edge [m]	-0.015325	-0.008414	-0.002511	-0.040683	0.016369	-0.016292	0.017856	0.022776	-0.010965	0.000501	-0.030647	-0.074262	25	0.301
dist_forest_edge ²	0.000074	0.000034	0.000021	0.000426	-0.000039	0.000065	-0.000252	-0.000358	0.000052	0.000001	0.000159	0.000217	53	0.301
canopy : conifers	-0.001392	0.001379	-0.001172	-0.000300	0.001586	0.000329	0.000733	0.000532	-0.001264	-0.001111	0.000369	0.000171	41	0.910
canopy : conifers ²	0.000020	-0.000026	0.000038	0.000005	-0.000018	-0.000005	-0.000024	0.000015	0.000297	0.000023	-0.000009	-0.000001	45	0.677
canopy : dist_road	-0.000201	0.000007	-0.000104	-0.000107	-0.000017	0.000197	-0.000279	-0.000053	0.000055	-0.000037	-0.000048	0.000489	28	0.424
canopy : dist_forest_edge	0.000371	-0.000322	0.000315	0.000476	-0.000131	0.000219	-0.000814	0.000068	0.000506	-0.000182	0.000249	0.000690	54	0.266
canopy:dist_forest_edge ²	-0.000002	0.000005	-0.000004	-0.000005	0.000000	-0.000001	0.000009	0.000002	-0.000011	0.000001	-0.000001	-0.000001	32	0.622

Table 10. Differences between the odds ratios of the day and night models. The difference was analysed using Wilcoxon rank sum test. A p-value < 0.05 indicates significance difference between day and night models concerning the particular model variable.

RD01	RD02	RD03	RD04	RD05	RD06	RD07	RD08	RD09	RD10	RD11	RD12	v	P-value
-0.003298	-0.029864	-0.020370	-0.007965	0.029595	0.027839	0.037520	0.000394	0.058724	-0.022897	-0.028936	-0.174640	36	0.850
0.100369	-0.466827	-1.066723	-0.312348	-0.041823	0.009513	-0.361136	-0.078179	0.447665	-0.291182	-0.284143	-0.653578	14	0.052
-0.000868	0.004012	0.014508	0.004326	-0.000525	-0.001094	0.002211	0.001494	-0.023860	0.003793	0.001228	0.004291	60	0.110
-0.072134	-0.035233	0.030056	-0.046693	-0.016865	0.046441	-0.038798	-0.086725	-0.121806	-0.111706	-0.119027	-0.036333	8	0.012
-0.036652	0.017623	-0.049146	-0.106050	-0.016100	-0.012952	-0.072645	-0.025505	0.022518	0.026145	-0.064351	0.134135	25	0.301
0.000525	-0.000345	0.000179	0.001368	0.000299	-0.000030	0.001004	0.000046	-0.000025	-0.000111	0.000296	-0.000980	53	0.301
-0.045785	0.053740	-0.168201	0.021653	-0.058692	-0.077909	0.168861	0.029426	0.150968	-0.036285	0.033493	-0.338098	32	0.622
0.000312	0.000036	0.002022	-0.000154	0.000260	0.000370	-0.001120	-0.000515	-0.000480	0.000443	-0.000141	0.001891	46	0.622
-0.001549	0.004124	0.008631	0.003475	0.000123	-0.000791	0.003001	0.000224	-0.006446	0.002442	0.002305	0.007173	61	0.092
0.000017	-0.000043	-0.000171	-0.000053	0.000009	0.000019	-0.000022	-0.000013	0.000346	-0.000041	-0.000010	-0.000057	22	0.204
-0.000331	0.000307	-0.000144	-0.000230	-0.000642	-0.000397	-0.000598	-0.000089	-0.000478	-0.000261	0.000033	-0.001660	7	0.009
0.000621	-0.000422	0.000628	-0.000452	0.000665	0.000574	-0.001819	-0.000908	-0.002012	0.000337	-0.000474	0.002998	39	1.000
-0.000004	-0.000002	-0.000013	0.000003	-0.000003	-0.000003	0.000012	0.000010	0.000007	-0.000005	0.000002	-0.000021	31	0.569
	RD01 -0.003298 0.100369 -0.000868 -0.072134 -0.036652 0.000525 -0.045785 0.000312 -0.001549 0.000017 -0.000331 0.000621 -0.000004	RD01 RD02 -0.003298 -0.029864 0.100369 -0.466827 -0.000868 0.004012 -0.072134 -0.035233 -0.036652 0.017623 0.000525 -0.000345 -0.045785 0.053740 0.000312 0.00036 -0.001549 0.004124 0.00031 0.000307 0.000621 -0.000422 -0.000004 -0.000002	RD01 RD02 RD03 -0.003298 -0.029864 -0.020370 0.100369 -0.466827 -1.066723 -0.0072134 -0.035233 0.030056 -0.036652 0.017623 -0.049146 0.000525 -0.000345 0.000179 -0.045785 0.053740 -0.168201 0.0000312 0.0004124 0.008631 0.000017 -0.000043 -0.000171 -0.000331 0.000307 -0.000144 0.000621 -0.000422 0.000628 -0.000004 -0.000002 -0.00013	RD01 RD02 RD03 RD04 -0.003298 -0.029864 -0.020370 -0.007965 0.100369 -0.466827 -1.066723 -0.312348 -0.000868 0.004012 0.014508 0.004326 -0.072134 -0.035233 0.030056 -0.046693 -0.036652 0.017623 -0.049146 -0.106050 0.000525 -0.000345 0.000179 0.001368 -0.045785 0.053740 -0.168201 0.021653 0.000312 0.00036 0.002022 -0.000154 -0.001549 0.004124 0.008631 0.003475 0.000017 -0.000043 -0.00171 -0.000053 -0.00031 0.000307 -0.00144 -0.00230 0.000621 -0.00002 -0.00013 0.000033	RD01 RD02 RD03 RD04 RD05 -0.003298 -0.029864 -0.020370 -0.007965 0.029595 0.100369 -0.466827 -1.066723 -0.312348 -0.041823 -0.000868 0.004012 0.014508 0.004326 -0.000525 -0.072134 -0.035233 0.030056 -0.046693 -0.016865 -0.036652 0.017623 -0.049146 -0.106050 -0.016100 0.000525 -0.000345 0.000179 0.001368 0.000299 -0.045785 0.053740 -0.168201 0.021653 -0.058692 0.000312 0.000036 0.002022 -0.000154 0.000260 -0.001549 0.004124 0.008631 0.003475 0.000123 0.000017 -0.000043 -0.000171 -0.000053 0.000009 -0.000331 0.00307 -0.000144 -0.000230 -0.000642 0.000621 -0.00002 -0.00013 0.000003 -0.000003	RD01 RD02 RD03 RD04 RD05 RD06 -0.003298 -0.029864 -0.020370 -0.007965 0.029595 0.027839 0.100369 -0.466827 -1.066723 -0.312348 -0.041823 0.009513 -0.000868 0.004012 0.014508 0.004326 -0.000525 -0.001094 -0.072134 -0.035233 0.030056 -0.046693 -0.016865 0.046441 -0.036652 0.017623 -0.049146 -0.106050 -0.016100 -0.012952 0.000525 -0.000345 0.000179 0.001368 0.000299 -0.000030 -0.045785 0.053740 -0.168201 0.021653 -0.058692 -0.077909 0.000312 0.000036 0.002022 -0.000154 0.000260 0.000370 -0.001549 0.004124 0.008631 0.003475 0.0000123 -0.000791 0.000017 -0.000043 -0.000171 -0.000053 0.0000642 -0.000397 0.000621 -0.0000422 0.000628 -0.0004	RD01 RD02 RD03 RD04 RD05 RD06 RD07 -0.003298 -0.029864 -0.020370 -0.007965 0.029595 0.027839 0.037520 0.100369 -0.466827 -1.066723 -0.312348 -0.041823 0.009513 -0.361136 -0.000868 0.004012 0.014508 0.004326 -0.000525 -0.01094 0.002211 -0.072134 -0.035233 0.30056 -0.046693 -0.016865 0.046441 -0.038798 -0.036652 0.017623 -0.0419146 -0.106050 -0.016100 -0.012952 -0.072645 0.000525 -0.000345 0.000179 0.001368 0.000299 -0.000030 0.001004 -0.045785 0.053740 -0.168201 0.021653 -0.058692 -0.077909 0.168861 0.000312 0.000036 0.002022 -0.000154 0.000260 0.000370 -0.001120 -0.0001549 0.004124 0.008631 0.003475 0.0000123 -0.000037 -0.000022 -0.000	RD01 RD02 RD03 RD04 RD05 RD06 RD07 RD08 -0.003298 -0.029864 -0.020370 -0.007965 0.029595 0.027839 0.037520 0.000394 0.100369 -0.466827 -1.066723 -0.312348 -0.041823 0.009513 -0.361136 -0.078179 -0.000868 0.004012 0.014508 0.004326 -0.000525 -0.001094 0.002211 0.001494 -0.072134 -0.035233 0.030056 -0.046693 -0.016100 -0.012952 -0.072645 -0.025505 -0.036652 0.017623 -0.049146 -0.106050 -0.016100 -0.012952 -0.072645 -0.025505 0.000525 -0.000345 0.000179 0.001368 0.000299 -0.000030 0.001004 0.000046 -0.045785 0.053740 -0.168201 0.021653 -0.058692 -0.077909 0.168861 0.029426 0.000312 0.000036 0.002022 -0.000154 0.000260 0.000370 -0.0001120 -0.000515 </td <td>RD01 RD02 RD03 RD04 RD05 RD06 RD07 RD08 RD09 -0.003298 -0.029864 -0.020370 -0.007965 0.029595 0.027839 0.037520 0.000394 0.058724 0.100369 -0.466827 -1.066723 -0.312348 -0.041823 0.009513 -0.361136 -0.078179 0.447665 -0.000868 0.004012 0.014508 0.004326 -0.000525 -0.001094 0.002211 0.001494 -0.023860 -0.072134 -0.035233 0.030056 -0.046693 -0.016100 -0.012952 -0.072645 -0.025505 0.022518 0.000525 -0.000345 0.000179 0.001368 0.000299 -0.000030 0.001004 0.000046 -0.000025 -0.045785 0.053740 -0.168201 0.021653 -0.058692 -0.077909 0.168861 0.029426 0.150968 0.000312 0.000036 0.002022 -0.000154 0.000260 0.000370 -0.000515 -0.000480 -0.001549 0.0</td> <td>RD01RD02RD03RD04RD05RD06RD07RD08RD09RD10-0.003298-0.029864-0.020370-0.0079650.0295950.0278390.0375200.0003940.058724-0.0228970.100369-0.466827-1.066723-0.312348-0.0418230.009513-0.361136-0.0781790.447665-0.291182-0.0008680.0040120.0145080.004326-0.000525-0.0010940.0022110.001494-0.0238600.003793-0.072134-0.0352330.030056-0.046693-0.0168650.046441-0.038798-0.086725-0.121806-0.111706-0.0366520.017623-0.049146-0.106050-0.016100-0.012952-0.072645-0.0255050.0225180.0261450.000525-0.0003450.0001790.0013680.000299-0.0003000.0010040.000046-0.000025-0.0001111-0.0457850.053740-0.1682010.021653-0.058692-0.0779090.1688610.0294260.150968-0.0362850.0003120.0000360.000202-0.0001540.0002600.000370-0.001120-0.000515-0.0004460.0024420.00017-0.00043-0.000171-0.000530.0000990.000019-0.000058-0.000478-0.0002610.000621-0.0004220.000628-0.0004520.0006650.000574-0.001819-0.000908-0.0020120.000377-0.000044-0.000033-0.000033-0.000033<!--</td--><td>RD01RD02RD03RD04RD05RD06RD07RD08RD09RD10RD11-0.003298-0.029864-0.020370-0.0079650.0295950.0278390.0375200.0003940.058724-0.022897-0.0289360.100369-0.466827-1.066723-0.312348-0.0418230.009513-0.361136-0.0781790.447665-0.291182-0.284143-0.0008680.0040120.0145080.004326-0.000525-0.0010940.0022110.001494-0.0238600.0037930.001228-0.072134-0.0352330.030056-0.046693-0.0168650.046441-0.038798-0.086725-0.121806-0.111706-0.119027-0.0366520.017623-0.049146-0.106050-0.016100-0.012952-0.072645-0.0255050.0225180.026145-0.0643510.000525-0.0003450.0001790.0013680.000299-0.0000300.0010040.00046-0.000025-0.0001110.000296-0.0457850.053740-0.1682010.021653-0.058692-0.0779090.1688610.0294260.150968-0.0362850.0334930.0003120.000360.002022-0.0001540.000173-0.0007910.0030010.000224-0.0064460.0024420.0023050.000017-0.00043-0.000171-0.000530.0000990.00019-0.000598-0.000478-0.0002610.000330.000021-0.0004220.000628-0.0006550.000574<</td><td>RD01 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-0.001094 0.002211 0.001494 -0.023860 0.003793 0.001228 0.004291 60 -0.072134 -0.035233 0.30056 -0.046693 -0.016905 -0.012952 -0.072645 -0.025505 0.02218 0.002118 0.064351 0.134135 25 0.000525 -0.000345 0.000179 0.001368 0.00299 -0.007090 0.168861 0.029426 0.150968 -0.0363493 -0.33898 32<</td></td>	RD01 RD02 RD03 RD04 RD05 RD06 RD07 RD08 RD09 -0.003298 -0.029864 -0.020370 -0.007965 0.029595 0.027839 0.037520 0.000394 0.058724 0.100369 -0.466827 -1.066723 -0.312348 -0.041823 0.009513 -0.361136 -0.078179 0.447665 -0.000868 0.004012 0.014508 0.004326 -0.000525 -0.001094 0.002211 0.001494 -0.023860 -0.072134 -0.035233 0.030056 -0.046693 -0.016100 -0.012952 -0.072645 -0.025505 0.022518 0.000525 -0.000345 0.000179 0.001368 0.000299 -0.000030 0.001004 0.000046 -0.000025 -0.045785 0.053740 -0.168201 0.021653 -0.058692 -0.077909 0.168861 0.029426 0.150968 0.000312 0.000036 0.002022 -0.000154 0.000260 0.000370 -0.000515 -0.000480 -0.001549 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<td>RD01RD02RD03RD04RD05RD06RD07RD08RD09RD10RD11-0.003298-0.029864-0.020370-0.0079650.0295950.0278390.0375200.0003940.058724-0.022897-0.0289360.100369-0.466827-1.066723-0.312348-0.0418230.009513-0.361136-0.0781790.447665-0.291182-0.284143-0.0008680.0040120.0145080.004326-0.000525-0.0010940.0022110.001494-0.0238600.0037930.001228-0.072134-0.0352330.030056-0.046693-0.0168650.046441-0.038798-0.086725-0.121806-0.111706-0.119027-0.0366520.017623-0.049146-0.106050-0.016100-0.012952-0.072645-0.0255050.0225180.026145-0.0643510.000525-0.0003450.0001790.0013680.000299-0.0000300.0010040.00046-0.000025-0.0001110.000296-0.0457850.053740-0.1682010.021653-0.058692-0.0779090.1688610.0294260.150968-0.0362850.0334930.0003120.000360.002022-0.0001540.000173-0.0007910.0030010.000224-0.0064460.0024420.0023050.000017-0.00043-0.000171-0.000530.0000990.00019-0.000598-0.000478-0.0002610.000330.000021-0.0004220.000628-0.0006550.000574<</td> <td>RD01 RD02 RD03 RD04 RD05 RD06 RD07 RD08 RD09 RD10 RD11 RD12 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32<</td>	RD01RD02RD03RD04RD05RD06RD07RD08RD09RD10RD11-0.003298-0.029864-0.020370-0.0079650.0295950.0278390.0375200.0003940.058724-0.022897-0.0289360.100369-0.466827-1.066723-0.312348-0.0418230.009513-0.361136-0.0781790.447665-0.291182-0.284143-0.0008680.0040120.0145080.004326-0.000525-0.0010940.0022110.001494-0.0238600.0037930.001228-0.072134-0.0352330.030056-0.046693-0.0168650.046441-0.038798-0.086725-0.121806-0.111706-0.119027-0.0366520.017623-0.049146-0.106050-0.016100-0.012952-0.072645-0.0255050.0225180.026145-0.0643510.000525-0.0003450.0001790.0013680.000299-0.0000300.0010040.00046-0.000025-0.0001110.000296-0.0457850.053740-0.1682010.021653-0.058692-0.0779090.1688610.0294260.150968-0.0362850.0334930.0003120.000360.002022-0.0001540.000173-0.0007910.0030010.000224-0.0064460.0024420.0023050.000017-0.00043-0.000171-0.000530.0000990.00019-0.000598-0.000478-0.0002610.000330.000021-0.0004220.000628-0.0006550.000574<	RD01 RD02 RD03 RD04 RD05 RD06 RD07 RD08 RD09 RD10 RD11 RD12 -0.003298 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-0.072134 -0.035233 0.30056 -0.046693 -0.016905 -0.012952 -0.072645 -0.025505 0.02218 0.002118 0.064351 0.134135 25 0.000525 -0.000345 0.000179 0.001368 0.00299 -0.007090 0.168861 0.029426 0.150968 -0.0363493 -0.33898 32<

Model variable	RD01	RD02	RD03	RD04	RD05	RD06	RD07	RD08	RD09	RD10	RD11	RD12
canopy [%]	0.362	0.013	0.154	0.938	0.003	0.003	0.200	0.451	0.012	0.650	0.067	0.026
conifers [%]	0.958	0.001	< 0.001	0.039	0.127	0.665	0.011	0.146	0.192	0.102	0.455	0.729
conifers ²	0.368	< 0.001	< 0.001	0.913	0.391	0.365	0.032	0.225	0.060	0.898	0.528	0.727
slope [°]	< 0.001	0.012	0.755	0.774	0.309	0.001	< 0.001	0.022	< 0.001	0.286	0.005	0.004
dist_road [m]	0.027	< 0.001	< 0.001	0.072	0.001	0.191	< 0.001	0.095	< 0.001	< 0.001	0.129	0.024
dist_road ²	0.084	0.087	0.005	0.227	0.009	0.921	< 0.001	0.059	0.039	< 0.001	0.962	< 0.001
dist_forest_edge [m]	0.444	0.765	0.594	< 0.001	0.018	0.001	0.014	< 0.001	0.009	0.758	< 0.001	0.056
dist_forest_edge ²	0.347	0.398	0.121	0.030	0.003	0.001	0.117	0.001	0.028	0.246	0.001	0.327
canopy : conifers	0.771	0.005	< 0.001	0.134	0.089	0.719	0.004	0.282	0.236	0.606	0.546	0.861
canopy : conifers ²	0.175	< 0.001	< 0.001	0.924	0.255	0.426	0.019	0.554	0.098	0.446	0.510	0.864
canopy : dist_road	0.888	0.029	0.727	0.554	0.102	0.243	0.009	0.339	< 0.001	0.984	0.940	0.534
canopy:dist_forest_edge	0.938	0.399	0.433	0.006	0.065	0.019	0.007	< 0.001	0.126	0.342	0.053	0.182
canopy : dist_forest_edge ²	0.854	0.653	0.175	0.062	0.020	0.035	0.012	< 0.001	0.563	0.108	0.026	0.663

 Table 11. P-values of the model variables in the individual weekend models.

 Table 12. P-values of the model variables in the individual workday models.

Model variable	RD01	RD02	RD03	RD04	RD05	RD06	RD07	RD08	RD09	RD10	RD11	RD12
canopy [%]	0.271	0.009	0.033	0.800	0.019	0.066	0.842	0.122	0.067	0.023	0.719	0.221
conifers [%]	0.006	0.017	< 0.001	0.039	0.375	0.367	0.016	0.516	0.204	0.007	0.546	0.723
conifers ²	0.007	0.029	< 0.001	0.750	0.087	0.079	0.047	0.342	0.078	0.054	0.984	0.795
slope [°]	< 0.001	0.001	0.794	0.013	0.355	0.036	< 0.001	< 0.001	< 0.001	0.001	0.648	< 0.001
dist_road [m]	0.003	< 0.001	< 0.001	0.001	< 0.001	0.309	< 0.001	0.003	0.001	< 0.001	0.020	0.831
dist_road ²	0.074	< 0.001	0.011	0.006	0.006	0.411	0.003	0.010	0.113	< 0.001	0.759	0.469
dist_forest_edge [m]	0.974	0.429	0.753	< 0.001	< 0.001	0.013	0.102	0.001	0.174	0.718	0.378	0.057
dist_forest_edge ²	0.786	0.287	0.180	< 0.001	< 0.001	0.022	0.596	0.052	0.332	0.258	0.407	0.278
canopy : conifers	0.029	0.087	< 0.001	0.104	0.285	0.332	0.009	0.429	0.131	0.022	0.834	0.707
canopy : conifers ²	0.040	0.072	< 0.001	0.869	0.072	0.092	0.025	0.208	0.062	0.108	0.898	0.740
canopy : dist_road	0.147	0.021	0.030	0.223	0.046	0.953	0.534	0.939	< 0.001	0.555	0.524	0.738
canopy : dist_forest_edge	0.291	0.907	0.538	< 0.001	0.005	0.145	0.408	< 0.001	0.772	0.673	0.954	0.283
canopy : dist_forest_edge ²	0.386	0.274	0.807	0.002	0.001	0.255	0.453	< 0.001	0.128	0.954	0.729	0.632

Model variable	RD01	RD02	RD03	RD04	RD05	RD06	RD07	RD08	RD09	RD10	RD11	RD12
canopy [%]	0.196	0.007	0.040	0.718	0.865	0.755	0.012	0.919	0.020	0.055	0.067	0.022
conifers [%]	0.345	< 0.001	< 0.001	< 0.001	0.654	0.937	< 0.001	0.152	0.934	< 0.001	< 0.001	0.005
conifers ²	0.440	< 0.001	< 0.001	0.101	0.345	0.214	0.002	0.154	0.361	0.003	0.007	0.006
slope [°]	< 0.001	< 0.001	0.720	0.626	0.842	0.154	< 0.001	0.321	< 0.001	0.137	0.001	< 0.001
dist_road [m]	0.001	0.017	< 0.001	< 0.001	0.018	0.303	< 0.001	< 0.001	0.058	0.214	< 0.001	0.657
dist_road ²	0.001	0.272	0.001	< 0.001	0.005	0.275	< 0.001	< 0.001	0.028	0.002	0.142	0.130
dist_forest_edge [m]	0.240	0.066	< 0.001	0.001	< 0.001	0.008	0.004	0.021	< 0.001	0.380	0.605	0.002
dist_forest_edge ²	0.150	0.895	< 0.001	0.012	< 0.001	0.003	0.376	0.013	0.004	0.131	0.777	0.004
canopy : conifers	0.639	< 0.001	< 0.001	0.001	0.915	0.600	0.001	0.418	0.340	0.004	0.001	0.002
canopy : conifers ²	0.981	< 0.001	< 0.001	0.127	0.284	0.139	0.005	0.480	0.165	0.022	0.017	0.003
canopy : dist_road	0.650	0.137	0.847	0.988	0.105	0.118	0.001	0.708	0.245	0.451	0.181	0.081
canopy : dist_forest_edge	0.571	0.188	0.098	0.002	0.004	0.147	0.015	< 0.001	0.001	0.471	0.197	0.014
canopy:dist_forest_edge ²	0.495	0.802	0.006	0.022	0.001	0.070	0.239	< 0.001	0.118	0.207	0.436	0.020

 Table 13. P-values of the model variables in the individual day models.

 Table 14. P-values of the model variables in the individual night models.

Model variable	RD01	RD02	RD03	RD04	RD05	RD06	RD07	RD08	RD09	RD10	RD11	RD12
canopy [%]	0.229	0.059	0.464	0.539	0.033	0.027	0.250	0.925	< 0.001	0.084	0.545	0.985
conifers [%]	< 0.001	0.354	0.105	0.468	0.733	0.723	0.026	0.538	0.198	0.260	0.035	0.079
conifers ²	0.002	0.267	0.113	0.466	0.878	0.609	0.031	0.819	0.340	0.972	0.061	0.369
slope [°]	0.004	0.047	0.285	0.024	0.918	0.017	< 0.001	< 0.001	0.011	< 0.001	< 0.001	< 0.001
dist_road [m]	0.034	< 0.001	0.001	0.093	0.048	0.431	< 0.001	0.017	< 0.001	< 0.001	0.746	0.004
dist_road ²	0.409	< 0.001	0.042	0.281	0.055	0.112	0.020	0.019	0.003	< 0.001	0.020	< 0.001
dist_forest_edge [m]	0.530	0.216	0.057	< 0.001	0.069	0.491	0.787	0.004	0.144	0.734	0.001	0.773
dist_forest_edge ²	0.530	0.949	0.340	0.003	0.046	0.564	0.999	0.282	0.290	0.371	0.002	0.231
canopy : conifers	0.001	0.361	0.270	0.970	0.735	0.518	0.009	0.585	0.329	0.631	0.080	0.050
canopy : conifers ²	0.004	0.199	0.162	0.312	0.836	0.462	0.011	0.834	0.376	0.645	0.065	0.265
canopy : dist_road	0.040	0.607	0.011	0.059	0.339	0.641	0.795	0.375	< 0.001	0.036	0.002	0.082
canopy : dist_forest_edge	0.163	0.264	0.662	0.001	0.246	0.593	0.245	< 0.001	0.103	0.829	0.044	0.507
canopy : dist_forest_edge ²	0.136	0.552	0.486	0.022	0.179	0.770	0.110	0.093	0.648	0.378	0.030	0.068

Table 15. The percentage of GPS locations which lay within the forest when using all locations (winter), locations from Saturday and Sunday (weekend), locations from Tuesday and Thursday (workday), locations from 1h after sunrise to 1h before sunset (day) or locations from 2h after sunset to 2h before sunrise (night). Statistics were conducted among roe deer which were exposed to intense recreational activities (peri-urban roe deer), among roe deer which were exposed to low recreational activities (rural roe deer) and among all roe deer (all).

Animal ID	Winter [%]	Weekend [%]	Workday [%]	Day [%]	Night [%]
RD01	94	93	95	98	89
RD02	86	85	86	97	80
RD03	74	75	76	92	60
RD04	84	83	85	94	76
RD05	99	100	99	100	98
RD06	98	97	100	100	96
RD07	74	76	74	86	65
RD08	79	81	79	97	64
RD09	59	62	56	90	34
RD10	78	78	77	100	61
RD11	80	80	80	99	64
RD12	97	98	98	100	94
mean: peri-urban	86	86	86	96	78
mean: rural	82	82	85	91	71
mean: all	84	84	84	96	73
SD: peri-urban	9	9	9	5	13
SD: rural	13	12	15	12	22
SD: all	12	11	12	4	18

P-values of the model variables

The p-values of the model variables in the model fitted with each the WE, the WD, the day and the night sample differed (WE: χ^2 = 22.00, df = 12, P = 0.038; WD: χ^2 = 13.12, df = 12, P = 0.027; day: χ^2 = 21.35, df = 12, P = 0.045; night: χ^2 = 24.66, df = 12, P = 0.017). However, Nemenyi post-hoc test revealed only tendencies within the WE models (for all model variables: P > 0.140), the nocturnal models (for all model variables: P > 0.140) and the diurnal models (for all model variables: P > 0.120). In contrast, on WD *dist_road* had a significant lower p-value than *canopy : dist_road* (P = 0.044) and *canopy : dist_forest_edge* (P = 0.044). This indicates that the variable *dist_road* was more important than *canopy : dist_road* and *canopy : dist_forest_edge* for explaining habitat selection on WD. Further, we compared the p-values of each model variable between the WE and WD models as well as between the diurnal and the nocturnal models. *Dist_forest_edge*² had lower p-values in the WE than in the WD models (V = 66, P = 0.034). Apart from that the p-values between WE and WD revealed no difference (for all other model variables: 24 < V < 63, P > 0.076). In contrast, the p-values of the following model variables were lower during the day than during the night: *conifers*²: V = 69, P = 0.016; *dist_forest_edge*: V = 65, P = 0.042; *dist_forest_edge*²: V = 66, P = 0.034; *canopy : conifers*²: V = 67, P = 0.027). The p-values of the remaining model variables did not difference (for all other model variables).

between day and night: *canopy*: V = 49, P = 0.470; *conifers*: V = 55, P = 0.233; *slope*: V = 30, P = 0.519; *dist_road*: V = 58, P = 0.151; *dist_road*²: V = 49, P = 0.470; *canopy : conifers*: V = 56, P = 0.204; *canopy : dist_road*: V = 36, P = 0.850; *canopy : dist_forest_edge*: V = 62, P = 0.077; *canopy : dist_forest_edge*²: V = 54, P = 0.266.

Variation in the use of open areas during the night

The use of open areas during the night had a high variance which we could neither explain with the classification peri-urban and rural roe deer (F-statistics = 0.42, P = 0.530), nor with the sex (F-statistics = 0.22, P = 0.649), the density of surfaced roads (F-statistics = 3.14, P = 0.107), winter HR size (F-statistics = 0.83, P = 0.385), the nocturnal HR size (F-statistic = 0.64, P = 0.444), or the overlap between diurnal and nocturnal CA (F-statistics = 0.01, P = 0.934). In contrast, the use of forest during the night was positively correlated with the forest proportion within the HR of the roe deer (F-statistics = 45.76, P < 0.001). However, this was expected as both, the use of forest and the forest proportion within the HR, depends on the same GPS locations. Further, roe deer spent significantly more time in open areas during the night if the density of gravel roads was low (F-statistics = 11.87, P = 0.006), which is also the case when using all winter (F-statistic = 7.71, P = 0.020, weekend (F-statistic = 8.63, P = 0.015) and workday locations (F-statistic = 7.93, P = 0.019). However, this was not the case during the day (F-statistic = 0.39, P = 0.54). Thus, roe deer selected forest independent of the density of gravel roads during the day. However, the density of gravel roads was positively correlated with the forest proportion within the HR (F-statistic = 6.67, P = 0.027). Additionally, the use of forest during the night was negatively correlated with mean slope (F-statistics = 10.58, P = 0.009), but also slope was significantly negative correlated with the forest proportion within the HR (F-statistics = 13.99, P = 0.004). Because of the correlations between density of gravel roads and forest proportion as well as between slope and forest proportion, we could not made any conclusions which factors influenced the time spent in open areas.

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Appendix IV: Autocorrelation

In general, analysis of GPS data is not biased due to autocorrelation if the individual animals rather than the locations are used as sample units (Garton et al. 2001). Regarding KDE, several authors (e.g. De Solla et al. 1999, Blundell et al. 2001, Fieberg 2007) found that increasing autocorrelated data improve the accuracy and precision of the home range estimates. Consequently, KDE do not require absolute temporal independence (Swihart et al. 1997, De Solla et al. 1999) and data should rather be equally spaced throughout the study period and be collected throughout the daily and seasonal period of interest (De Solla et al. 1999, Kernohan et al. 2001, Fieberg 2007, Silvy 2012). Further, in habitat selection analysis following the method NB RSF, temporal autocorrelation of the locations creates no bias because relative counts are modelled (Nielson et al. 2013).

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Appendix V: prediction of food supply and concealment cover

Concealment cover and forage availability are along with thermal cover important habitat variables which influence habitat selection of roe deer (Tufto et al. 1996, Ewald et al. 2014) and may occur as a trade-off (Mysterud et al. 1999, Borkowski et al. 2008). Thus, we intended to build a model which estimates concealment cover and forage availability depending on environmental factors such as canopy cover or solar radiation. With this model we would have estimated concealment cover and forage availability for each circle unit in the NB RSF approach. Thus, we would have been able to disentangle the effect of habitat quality (concealment cover and food supply) from that of human disturbance. For this purpose, concealment cover and forage availability of various forest types were recorded in the field (see recording sheet on the compact disk). The field work was conducted in September and October by two different persons. To ensure similar recording, the two persons recorded the measurements together during one day. We developed the sampling design of the field recording in ArcMap 10.1 (ESRI, Redlands, CA, USA) and in the software R 3.1.1 (R Core Team 2014). First, we grouped the forested area into three forest types according to their canopy cover, conifer proportion and development stage. Canopy cover was classified into (1) 0-33%, (2) 34-66% and (3) 67-100%; conifer proportion into (1) 90-100% , (2) 11-89% and (3) 0-10% conifers; development stage into (1) regeneration and thickening (diameter at breast height, DBH < 12 cm), (2) pole wood (DBH 12-30 cm), (3) tree wood (DBH > 30 cm) and (4) heterogeneous forest stands. Each forest type was an unique combination of this habitat variables, leading to 21 different forest types in our study area. Next, we laid a regular point grid with 50x50 m within the forest of the study area, whereof we randomly selected 287 out of the 2327 points for the field recording. All forest types were selected at least ones and then according to their frequency. The selected points were at least 20 m apart from the edge of the next different forest type as well as from the roads and the forest edge. Next, we recorded concealment cover and forage availability in the field. Both variables had a index reaching from one to five, where five meant low forage availability or high concealment cover. The forage availability index (FAI; see also recording sheet on the compact disk) is based on the cover of different plant species groups: Rubus sectio Rubus, Rubus sp., attractive conifers, remaining conifers, deciduous trees and shrubs as well as herbs and grasses. The sum of the covers of different species groups could exceed 100%. We assigned the specific index to the site if one of the characteristics in the classification was fulfilled (e.g. FAI = 3 if 5% attractive conifers or 10 to 20% Rubus sectio Rubus). If the characteristics in several classes were fulfilled, we assigned the lower index, which means the higher forage availability to the site. The covers were measured on a square with 310 m². As the the diet composition of roe deer differs between summer and winter (Tixier et al. 1997, Barancekova et al. 2010), we built a index each for summer (FAIS) and winter (FAIW). However, roe deer select Rubus sp. over all seasons (Klötzli 1965, Tixier et al. 1996, Cornelis et al. 1999, Moser et al. 2006,

Moser et al. 2008). Thus, a high cover of Rubus sp. led to a low forage availability index - which means high food supply - in both seasons. In contrast, conifers are mainly preferred in winter (Klötzli 1965, Cornelis et al. 1999). However, Moser et al. (2006) found that roe deer also select conifers in summer, although, conifers have relatively low nutrient values during this period. Thus, conifers could led to a low forage availability index in winter, but only to a medium food supply in summer if they occurred frequently. We differentiated attractive conifers: European silver fir (Abies alba), English yew (Taxus baccata) and Douglas fir (Pseudotsuga menziesii; Kupferschmid et al. 2010) from the remaining conifer tress (e.g. Norway spruce [Picea abies]). Broadleaved trees and shrubs are mainly preferred during summer, but also foraged in winter (Klötzli 1965, Cornelis et al. 1999, Moser et al. 2006, Moser et al. 2008). Thus, they could led to a low forage availability index in summer, but only to a medium food supply in winter if they occurred frequently. Herbs are used in summer but not as much as broadleaved trees, shrubs and Rubus sp. (Cornelis et al. 1999, Moser et al. 2006). Shrubs and coniferous trees did not include English holly (Ilex aquifolium). Further, we did not include any spiny herbs in the group herbs. As forage is only accessible to roe deer under 1.20 m (Duncan et al. 1998, Pettorelli et al. 2001), we only considered plants below this value. Concealment cover index (CCI) was measured with a cover pole, described by Griffith et al. (1988) and thus included vegetation and topography. Measurements were taken in the standing positions (eye level: about 1.60 m), as we studied the influence of human disturbance on the roe deer. Further, we took the measurements in the four cardinal directions with one eye closed (recommended by Collins et al. 2001) and expressed them by concealment percentage for the four sections: 0 to 50 cm, 50 to 100 cm, 100 to 150 cm, 150 to 200 cm (see Griffith et al. 1988). Based on these results, we calculated the weighted concealment according to Griffith et al. (1988).

We extracted the following habitat variables for each sample unit in R 3.1.1 (R Core Team 2014): *canopy cover, conifer proportion, slope, aspect, development stage, total radiation, direct radiation, diffuse radiation, distance to road, distance to forest edge* and *distance to edge*. We included roads, forest edges, streams and edges of the forest types in the variable *distance to edge*. We tried to model the concealment cover as well as the forage availability as a function of these habitat variables using linear models as well as generalized linear models (negative binomial, Poisson, gamma and beta distribution). However, the coefficient of determination R² did not exceed 0.30 and was usually around 0.10. A principal components analysis and a cluster analysis revealed that no pattern within the data exists. Thus, we decided to model the habitat selection of the roe deer with the habitat variables *canopy, conifers, slope, dist_road, dist_build* and *dist_forest_edge*.

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