#### ORIGINAL ARTICLE



# Habitat suitability vs landscape connectivity determining roadkill risk at a regional scale: a case study on European badger (Meles meles)

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#### Abstract

Collisions between wildlife and vehicles represent the main conflict between infrastructures and ecosystems. Road mortality is the largest single cause of death for many vertebrates, representing a growing phenomenon of remarkable dimension. Most studies in road ecology investigated spatial roadkill patterns, showing that roadkill probability is often higher near optimal habitat for a large amount of species. Landscape connectivity has been less often considered in roadkill research, and only few studies considered habitat suitability and landscape connectivity at the same time. The purpose of the present study was to evaluate the relative importance of habitat suitability and landscape connectivity in determining roadkill risk for a habitat-generalist carnivore, namely, the Eurasian badger in the Abruzzo region (Central Italy). We collected occurrence data of living individuals from camera trapping and roadkill data of through a Citizen Science initiative. We used the occurrence data to produce a habitat suitability model (HSM) and a landscape connectivity model (LCM). Both HSM and LCM were then used as predictors in combination with road characteristics to fit a roadkill risk model. We found that landscape connectivity was more important than habitat suitability in determining roadkill risk for the Eurasian badger. Overall, the density of regional roads was the most important variable. Our finding highlighted how important is to consider landscape connectivity in planning mitigation measures aimed to preserve habitat-generalist species.

Keywords Habitat generalist · Habitat suitability · Landscape connectivity · Roadkill · Eurasian badger Meles meles

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# Introduction

Human-wildlife interactions are constantly increasing worldwide, with relevant impacts on biodiversity and ecosystems (Sala et al. [2000;](#page-8-0) Sinclair and Byrom [2006](#page-8-0)). Road networks are known to provide a noticeable contribution to human footprint (Sanderson et al. [2002;](#page-8-0) Ibisch et al. [2016\)](#page-7-0). The impacts of roads on wildlife include habitat loss and fragmentation; chemical, light, and noise pollution; barrier effect; spread of invasive species; and mortality by roadkill (Forman et al. [2003;](#page-7-0) Van der Ree et al. [2015\)](#page-8-0). In particular, wildlife-vehicle collisions are among the most common road-related impacts on animal populations (Forman and Alexander [1998](#page-7-0); Coffin [2007\)](#page-7-0), causing great economic damages and entailing serious risks for driver safety (Conover et al. [1995](#page-7-0); Seiler [2005](#page-8-0)).

A very large number of vertebrates die worldwide along such linear infrastructures (González-Gallina et al. [2013;](#page-7-0) D'Amico et al. [2015](#page-7-0)), sometimes with significant impacts on the population persistence of species (Fahrig et al. [1995;](#page-7-0) Mumme et al. [2000;](#page-8-0) Borda-de-Agua et al. [2014\)](#page-7-0). Consequently, in the last decades, the number of studies

focusing on the factors determining roadkill risk has increased considerably (Forman et al. [2003;](#page-7-0) Van der Ree et al. [2015](#page-8-0); D'Amico et al. [2018\)](#page-7-0). A central topic in roadkill research is the spatial distribution of casualties, which is probably the most investigated issue in road ecology (e.g., Van der Ree et al. [2015](#page-8-0); D'Amico et al. [2018\)](#page-7-0). Most studies have investigated the association between roadkill probability and some spatial variables characterizing road features or traffic volume, but also environmental predictors are increasingly considered in roadkill research (Litvaitis and Tash [2008;](#page-8-0) D'Amico et al. [2015](#page-7-0)). Unsurprisingly, several studies focusing on a large amount of different species showed that roadkill probability is higher near optimal habitat for such species (Roger et al. [2012;](#page-8-0) D'Amico et al. [2015;](#page-7-0) Visintin et al. [2016\)](#page-9-0). Landscape connectivity, defined as the degree to which the landscape facilitates or impedes individual movement among patches of resources (Taylor et al. [1993\)](#page-8-0), has been less often considered in roadkill research, whereas only few studies considered habitat suitability and landscape connectivity at the same time. Some of them considered landscape connectivity as a direct consequence of habitat suitability, i.e., the distance between suitable patches of habitat for given species (Grilo et al. [2011](#page-7-0); Kang et al. [2016\)](#page-8-0). Such studies showed that the roadkill risk of forest mammals was higher in well-connected areas (Grilo et al. [2011;](#page-7-0) Santos et al. [2013](#page-8-0); Kang et al. [2016](#page-8-0)), but their approach was inadequate to disentangle habitat suitability from landscape connectivity. Among the studies that included both habitat suitability and landscape connectivity in their analyses of roadkill risk, some were focused on species highly dependent on given habitats, thus highlighting habitat suitability as more important than landscape connectivity (Girardet et al. [2015](#page-7-0)). Such evidence suggests that both habitat suitability and landscape connectivity might be relevant factors in determining roadkill risk, although their relative importance depends on the habitat specialization degree of the analyzed species (Girardet et al. [2015](#page-7-0)). Moving beyond the studies that explored this pattern for habitat specialists (e.g., Santos et al. [2013;](#page-8-0) Girardet et al. [2015\)](#page-7-0), we aimed to evaluate the relative importance of habitat suitability and landscape connectivity for a habitat-generalist carnivore, i.e., the Eurasian badger (Meles meles; Kruuk [1989;](#page-8-0) De Marinis et al. [2002;](#page-7-0) Virgós et al. [2005](#page-9-0)), one of the most road-killed carnivores in Europe (e.g., Seiler et al. [2004;](#page-8-0) Jaarsma et al. [2007](#page-8-0); Grilo et al. [2009](#page-7-0)). Particularly, the low degree of specialization of the badger makes the species able to exploit small patches of natural or semi-natural habitats, rather than strictly relying on wide, contiguous areas (Dondina et al. [2018](#page-7-0)). Consequently, we hypothesized that landscape connectivity will be more important than habitat suitability in determining the roadkill risk of this species. Since some previous studies on this topic were carried out at a local scale (Santos et al. [2013;](#page-8-0) Kang et al. [2016\)](#page-8-0), we also decided to establish our study area at a regional scale, focusing on the Abruzzo region in Central Italy.

## **Methods**

#### Study area

The study was set in the Abruzzo region (Central Italy; Fig. [1\)](#page-2-0). The region covers ca.  $10,800 \text{ km}^2$ , with an elevation ranging from the sea level to ca. 2900 m a.s.l. The climate of the region is Mediterranean along the coast (with hot/dry summers and mild/rainy winters) and Continental in the inner Apennine mountain areas (with warm/dry summers and cold winters with abundant rain and snow). This region hosts several ecosystems: the coastal plains include mainly urbanized and cultivated environments with fragmented natural habitats (especially along riparian strips). Mixed forests are located throughout the sub-montane areas, while beech forests and montane shrublands and grasslands occur towards the higher elevations. The region is totally included in the distribution range of the badger in Italy (Rondinini et al. [2013](#page-8-0); Kranz et al. [2016\)](#page-8-0). The human population density is ca. 122 inhabitants/ km2 (higher in the coastal plains and decreasing towards the inner mountains). The total length of the road network is 21,249.46 km, and the road density is  $1.96$  km/km<sup>2</sup> (70% local roads, 27% regional roads, and 2.6% highways; OpenStreetMap, [https://planet.openstreetmap.org.](https://planet.openstreetmap.org) Accessed December 2017).

#### Species data

We opportunistically collected occurrence data of both living and road-killed badgers. The occurrence data about living badgers were originally gathered by camera trapping carried out between 2013 and 2016 and made available upon request by a number of protected areas in the study region (i.e., Majella National Park, Monte Genzana Alto Gizio Nature Reserve, Zompo Lo Schioppo Nature Reserve, Calanchi di Atri Nature Reserve, Castel Cerreto Nature Reserve, and Cascate del Verde Nature Reserve). The roadkill data were collected between 2007 and 2016 along the whole regional road network through a Citizen Science initiative (Centro Studi per le Reti Ecologiche; [csre.riservagenzana.it](http://csre.riservagenzana.it)). Although opportunistic data may provide accurate predictions of species distribution (Tiago et al. [2017](#page-8-0)), they are often spatially auto-correlated and/or discontinuous (Boitani et al. [2011\)](#page-7-0) due to a likely unbalanced sampling effort that can vary widely across space (van Strien et al. [2013\)](#page-9-0). Therefore, both the living and the roadkill occurrences were checked for spatial aggregation (Suppl. Mat. 1). After this filtering step, we obtained 57 records for living badgers and 39 for roadkill events (Fig. [1](#page-2-0); Tables S1 and S2). Since living and roadkill occurrences were gathered through an opportunistic sampling, we tested if these sampled data were able to comprehensively capture the environmental variability represented in the study area (Suppl. Mat. 4) (Fig. [2\)](#page-3-0).

<span id="page-2-0"></span>

Fig. 1 Study area. Blue diamonds indicate living badgers' occurrences, while yellow triangles locate roadkill events. Occurrences showed in the figure were already filtered for spatial aggregation. The red lines represent

the road network. A colored version of this figure is available in the online version of the paper

## Analytical framework

As a first step, we used the occurrence data about living badgers to produce a habitat suitability model (HSM) and a landscape connectivity model (LCM). Both HSM and LCM were then used as predictors in combination with road characteristics to fit a roadkill risk model (RRM) based on the roadkill occurrences (see below).

### Habitat suitability model

We calibrated the HSM using three topographic and two vegetation predictors: altitude, slope, topographic roughness, vegetative normalized difference vegetation index (NDVI), and non-vegetative NDVI. We selected these predictors according to the ecological requirements of the badger in Italy (e.g., Prigioni and Deflorian [2005;](#page-8-0) Prigioni et al. [2008;](#page-8-0) Balestrieri et al. [2009;](#page-7-0) Biancardi et al. [2014;](#page-7-0) Chiatante et al. [2017](#page-7-0)). Specifically, we selected altitude and slope as they proved particularly important in driving badger occurrence (i.e., the species mostly occurs from 1300 to 1500 m and between 21 and 40°; see Prigioni and Deflorian [2005](#page-8-0)). Furthermore, the role of the topographical roughness in determining badger occurrence was highlighted in Newton-Cross et al. [\(2007](#page-8-0)). As regards NDVI, we selected this index as it describes vegetation amount and it is strictly related to vegetation productivity. Indeed, the European badger shows a strong dependence on forests with high tree coverage, as well as of forest remnants, which offer shelter, setts, and food resources (fruits, earthworms, and insects; for the use of NDVI as a variable for predicting the distribution of the European badger, see also Requena-Mullor et al. [2014,](#page-8-0) [2017](#page-8-0); Santos et al. [2016\)](#page-8-0).

The three topographic predictors were obtained from a digital elevation model (DEM; data available from the Abruzzo Region Cartographic Service; Wilson et al. [2007\)](#page-9-0), while the two vegetation predictors were derived from the NDVI, a remotely sensed measure of greenness that is correlated to

<span id="page-3-0"></span>

Fig. 2 Roadkill risk map. Risk increases (decreases) towards red (blue) colors. A colored version of this figure is available in the online version of the paper

net primary productivity (Ramesh et al. [2017](#page-8-0)). We calculated the NDVI values from Landsat 7 imagery (data available from the U.S. Geological Survey) for the same years of the records of the living badgers. We obtained the NDVI values for the growing period by averaging the NDVI values for the vegetative period (April–September) and for the non-growing period (October–March) in order to account for a possible variation in habitat use (e.g., Mateo-Sánchez et al. [2015;](#page-8-0) Ramesh et al. [2017\)](#page-8-0). The five predictors were rasterized at a spatial resolution of 40 m. We finally checked the absence of multicollinearity among the predictors by posing a variance inflation factor  $\leq$  5 (Zuur et al. [2010\)](#page-9-0).

The Eurasian badger is a habitat generalist widely distributed across the Western Palaearctic (Kranz et al. [2016](#page-8-0)), the study area representing a small portion. Several evidences showed that the environmental truncation in the niche estimation obtained when the study area encompasses just a small portion of the species global range produces severely biased predictions (Barbet-Massin et al. [2010](#page-7-0); Raes [2012;](#page-8-0) Guisan et al. [2014\)](#page-7-0). Therefore, we used a hierarchical structure to produce the HSM. The model was first fitted considering the badger global range and bioclimatic variables (global HSM, further details were provided in Suppl. Mat. 2). The five local predictors described above were then used to refine the projections at a regional scale (Pearson et al. [2004](#page-8-0); Lomba et al. [2010;](#page-8-0) Gallien et al. [2012;](#page-7-0) Di Febbraro et al. [2015\)](#page-7-0). For HSM calibration, we randomly placed a set of 10,000 background points in the study area (Phillips et al. [2006](#page-8-0); Barbet-Massin et al. [2012\)](#page-7-0). HSM was calibrated using an ensemble forecasting approach, as implemented in the R package biomod2 (Thuiller et al. [2009\)](#page-8-0). We considered the following six modeling algorithms: generalized linear models (GLM), generalized additive models (GAM), generalized boosted models (GBM), random forests (RF), and maximum entropy models (MAXENT; Thuiller et al. [2009](#page-8-0)). The occurrence dataset was randomly split into a 70% sample, used for the calibration of the model, and a remaining 30%, used to evaluate the model predictive performance, repeating the procedure 20 times and averaging the results. The predictive performance of the model was assessed by measuring the area under the receiver operating characteristic curve (AUC; Hanley and McNeil [1982\)](#page-7-0) and the true skill statistic (TSS; Allouche et al. [2006\)](#page-7-0). According to AUC, prediction accuracy can be considered excellent (AUC > 0.90), good (0.80 > AUC < 0.90), fair  $(0.70 > \text{AUC} < 0.80)$ , and poor (AUC  $< 0.60$ ; Swets [1988](#page-8-0)). According to TSS, prediction accuracy can be considered excellent (TSS  $> 0.75$ ), good (0.40  $> AUC < 0.75$ ), and poor (TSS < 0.40; Landis and Koch [1977\)](#page-8-0). To avoid using poorly calibrated models, only projections from models with  $AUC \geq$ 0.70 were considered in all the subsequent analyses (Di Febbraro et al. [2016\)](#page-7-0). The model averaging was performed by weighting the individual model projections respectively by their AUC scores and averaging the results (Marmion et al. [2009](#page-8-0)). The final consensus HSM was projected over the Abruzzo region.

#### Landscape connectivity model

Landscape connectivity for badger in the study area was built using the CIRCUITSCAPE software (McRae et al. [2008](#page-8-0)). This software relies on the electrical circuit theory to incorporate multiple random walk pathways, working on a set of habitat nodes and a resistance surface to calculate the relative cost of moving through the entire landscape (McRae et al. [2008\)](#page-8-0). This approach is especially suitable for mesocarnivores, because it assumes that individuals have no inherent knowledge of the landscape beyond their immediate surroundings (McClure et al. [2016;](#page-8-0) Reed et al. [2017\)](#page-8-0). To avoid pitfalls related to wide area generalization (Pelletier et al. [2014](#page-8-0)) and arbitrary choice of source (start nodes) and destination (end nodes; Koen et al. [2014\)](#page-8-0), we adopted a "tiling" approach (Anderson et al. [2014;](#page-7-0) further details were provided in Suppl. Mat. 3). To produce the resistance surface used to build the LCM, we reclassified a land use/land cover map (data available from the Abruzzo Region Cartographic Service) into different landscape permeability values (Table S1), relying on species expert knowledge (Boitani et al. [2002,](#page-7-0) [2004](#page-7-0); Roscioni et al. [2014](#page-8-0)). The reclassified land use/land cover map was then rasterized at a 40-m spatial resolution.

#### The roadkill risk model

The roadkill risk model was calibrated following an ensemble forecasting procedure similar to the one previously used for HSM. We considered the maps derived from HSM and LCM as biological predictors. Moreover, since badger has been shown to avoid both traveled roads and urban settlements (Revilla et al. [2001;](#page-8-0) Frantz et al. [2010;](#page-7-0) Spinozzi et al. [2012\)](#page-8-0), as well as suffer high road mortality along minor and isolated roads (Clarke et al. [1998;](#page-7-0) Grilo et al. [2009;](#page-7-0) van Langevelde et al. [2009\)](#page-9-0), we also included distance from urban areas and roads as factors potentially affecting the roadkill risk for this species. Accordingly, we added the following predictors: the Euclidean distance from urban areas, the density of local roads, the density of regional roads,

and the density of highways. The three road categories represent a proxy of road width and, consequently, of traffic volume: low traffic for local roads, medium traffic for regional roads, and high traffic for highways (Jaeger et al. [2005](#page-8-0); D'Amico et al. [2015\)](#page-7-0). Therefore, local roads were considered easier barriers to cross than regional roads and highways, and regional roads easier than highways. The density of each road category was calculated through a moving window in the ESRI ArcGIS® software package. We considered a window radius of 750 m, according to the average home range size of badgers (Balestrieri et al. [2016;](#page-7-0) Kauhala and Holmala [2011;](#page-8-0) Molina-Vacas et al. [2009;](#page-8-0) Gaughran et al. [2018\)](#page-7-0). All the predictors were rasterized at a 40-m spatial resolution. We checked the absence of multicollinearity among the predictors by posing a variance inflation factor  $\leq$  5 (Zuur et al. [2010\)](#page-9-0) and placed a set of 10,000 background points in the study area (Phillips et al. [2006;](#page-8-0) Barbet-Massin et al. [2012](#page-7-0)). We constrained the points to fall within a buffer of 750-m radius around the road network. All the modeling settings used for HSM remained unchanged (Fig. [3](#page-5-0)).

## **Results**

## Roadkill risk

The roadkill risk map predicted through our model is presented in Fig. [2](#page-3-0). In general, higher values of risk are located in hillside areas and in valleys with large urban surfaces.

Global HSM and HSM at regional scale showed fair-togood levels of predictive performance, with AUC values of  $0.809 \pm 0.005$  and  $0.752 \pm 0.036$ , respectively, and TSS values of  $0.532 \pm 0.009$  and  $0.440 \pm 0.064$ , respectively. Also RRM reached good predictive performance scores, showing an AUC of  $0.804 \pm 0.012$  and a TSS of  $0.554 \pm 0.044$ . HSM at regional scale showed that most of the suitable areas for badgers occur along the foothills and the river valleys characterized by wooded and shrubby vegetation. LCM highlighted the intermontane areas and, to a lesser extent, the hilly areas to host several connectivity corridors for the species.

Focusing on our hypothesis, RRM variable importance showed that habitat suitability was less important than landscape connectivity in determining roadkill risk (0.11 vs 0.44, Fig. [3](#page-5-0)), while the density of regional roads was the most important predictor (0.48; Fig. [4\)](#page-5-0). In particular, roadkill risk was positively related with habitat suitability up to a tipping point (ca. 95% roadkill probability) at low-medium suitability values. On the other hand, a bell-shaped relationship emerged between roadkill risk and landscape connectivity, with a peak (95% roadkill probability) in correspondence of intermediate connectivity values (Fig. [3](#page-5-0)). As regards the other predictors, roadkill risk was directly related to regional and local road densities, while exhibiting poor or no relationship with the distance from urban areas and the density of highways.

#### <span id="page-5-0"></span>Fig. 3 Variable importance



# **Discussion**

We showed that both habitat suitability and landscape connectivity represent relevant factors in determining the roadkill risk for a habitat-generalist species like the Eurasian badger. Specifically, apart from the evidence that the density of regional roads was the most important predictor of risk, our results confirmed our initial hypothesis that landscape connectivity was more important than habitat suitability in determining the badger roadkill risk at a regional scale. Overall, the present study points out the often neglected relevance of landscape connectivity in roadkill studies, with significant implications in the planning of mitigation measures.



Fig. 4 Response curves describing the shape of the relationship between roadkill risk (y-axis) and values of the explanatory variables (x-axis). Each curve represents one variable

The relevance of landscape connectivity in determining roadkill risk was also highlighted by Grilo et al. [\(2011](#page-7-0)), Santos et al. [\(2013\)](#page-8-0), and Kang et al. ([2016](#page-8-0)). However, some of these studies considered landscape connectivity as a direct consequence of habitat suitability (i.e., the distance between suitable patches of habitat for given species; Grilo et al. [2011](#page-7-0); Kang et al. [2016](#page-8-0)) and were therefore unable to disentangle the role of habitat suitability from landscape connectivity. The only study, to our knowledge, considering at the same time habitat suitability and landscape connectivity in a roadkill risk analysis was carried out in France on the roe deer Capreolus capreolus, leading to opposite conclusions to our findings (Girardet et al. [2015\)](#page-7-0). Such study, indeed, showed that habitat suitability was more important than landscape connectivity in determining the roadkill risk of this species (Girardet et al. [2015\)](#page-7-0). However, such study was carried out in a rural landscape characterized by forest patches in a meadow/plantation matrix (Girardet et al. [2015\)](#page-7-0). In such context, forest patches represent an essential habitat for the roe deer, and for this reason, habitat availability was more important than landscape connectivity in determining roadkill hotspots (Girardet et al. [2015\)](#page-7-0). Our approach, focusing on a habitat-generalist species, confirmed an expectable, although partial, importance of habitat suitability in determining roadkill risk, and above all a more considerable relevance of landscape connectivity. Such findings are providing novel points of view in a long-standing debate in road-mortality studies, highlighting that landscape connectivity should be seriously accounted for when the target species is a habitat-generalist species not so dependent on the suitability of a given habitat. In fact, habitat-generalist species, such as the Eurasian badger in Italy, are widespread in a variety of environments (Prigioni and Deflorian [2005](#page-8-0); Prigioni et al. [2008;](#page-8-0) Balestrieri et al. [2009](#page-7-0); Biancardi et al. [2014](#page-7-0); Chiatante et al. [2017](#page-7-0)), and their roadkill risk is likely more linked to dispersal across the landscape rather than to territorial use of more suitable habitat patches, just as observed in other Mediterranean countries (Grilo et al. [2009](#page-7-0)). However, the availability of density estimates for badgers would represent a useful source of information to further support this hypothesis.

Interpreting the roadkill risk curves can help to further explore the actual influence of habitat suitability and landscape connectivity on casualty hazard. Although we focused on a habitat-generalist species, its roadkill risk increased according to the increase of habitat suitability, just as previously described in road-mortality literature in many species (e.g., Roger et al. [2012](#page-8-0); D'Amico et al. [2015;](#page-7-0) Visintin et al. [2016](#page-9-0)), even in studies taking into account landscape connectivity (Grilo et al. [2011;](#page-7-0) Santos et al. [2013;](#page-8-0) Kang et al. [2016\)](#page-8-0). Nevertheless, this roadkill risk curve decreased in correspondence with the highest habitat suitability. This could confirm the highest risk related to exploratory behavior across less suitable habitats (Grilo et al. [2009](#page-7-0)). Roadkill risk was also highest at intermediate values of landscape connectivity and negligible at both lowest and highest values. These findings confirm that low landscape connectivity impedes individual movement across the environmental matrix (Taylor et al. [1993](#page-8-0)). Nevertheless, these patterns might depend on the context where they were investigated, due to the high number of factors usually involved in determining habitat suitability and landscape connectivity. Overall, the areas with the highest roadkill risk for the habitat-generalist Eurasian badger were characterized by intermediate values of both habitat suitability and (especially) landscape connectivity. Such areas are especially suitable for dispersing individuals, which are usually described as the most susceptible to roadkill risk (Grilo et al. [2009\)](#page-7-0).

Road-related variables also played a relevant role in determining roadkill risk, as previously showed by the available literature, including the studies considering habitat suitability and landscape connectivity (i.e., road sinuosity in Grilo et al. [2011](#page-7-0); a generic roadkill index in Santos et al. [2013;](#page-8-0) road width and sinuosity and also distance from a crossing structure in Girardet et al. [2015;](#page-7-0) road length and slope and also traffic volume in Kang et al. [2016](#page-8-0)). In our case study, the most important road-related variable, slightly more relevant than landscape connectivity, was the density of regional roads. Such roads were the medium-sized infrastructures in our road network, entailing intermediate levels of traffic volume and allowed vehicle speed. These findings agree with the available road-mortality literature on Eurasian badgers, confirming that this species tends to avoid major roads and can suffer high roadkill rates in correspondence with minor and isolated roads (Clarke et al. [1998](#page-7-0); Grilo et al. [2009](#page-7-0); van Langevelde et al. [2009\)](#page-9-0).

Although the scale of our approach makes it difficult to plan small-scale mitigation measures (e.g., where to exactly build a wildlife road-crossing structure), the present study provides a useful example pointing out the importance of planning mitigation measures not only according to the habitat suitability of target species, but also where landscape connectivity promotes individual movements, especially for habitat-generalist species. This approach allows to identify the road sections with high roadkill risk at a regional scale, claiming further finer scale efforts to plan efficient mitigation measures.

The identification of landscape connectivity as a main factor in determining the roadkill risk of habitat-generalist species at a regional scale provides a decisive contribution to a long-standing debate in road-mortality studies. Therefore, we can state now that landscape connectivity should be surely taken into account for the planning of mitigation measures aimed to preserve species, especially the habitat-generalist ones, and to ensure driver safety.

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