



Ecological niche modelling of *Babesia* spp. infection in wildlife experimentally evaluated in northern Italy with reference to questing *Ixodes ricinus* ticks

Stefania Zanet, Ezio Ferroglio, Elena Battisti, Paolo Tizzani

Department of Veterinary Sciences, Università degli Studi di Torino, Grugliasco (TO), Italy

Abstract

Tick-borne diseases and especially those caused by protozoa of the genus *Babesia*, are gaining increasing attention as emerging zoonotic pathogens. Zoonotic species like *B. venatorum* and *B. microti* have wild animals as main reservoir hosts. We propose a habitat suitability model for *Babesia* spp. to better understand the entity of *Babesia* presence, to improve diagnostic awareness and to optimize screening and preventive actions. The probability of *Babesia* spp. presence was estimated by investigation of potential correlation between this protozoa in wild ruminants and the environmental factors that can favour or limit vector and host availability. We developed two separate models to evaluate the separate roles of cervids and alpine chamois for *Babesia* spp. epidemiology. A comprehensive model using all presence data from all ungu-

Correspondence: Stefania Zanet, Department of Veterinary Sciences, Università degli Studi di Torino, Largo Braccini 2, 10095 Grugliasco (TO), Italy Tel.: +39 0116708442

E-mail: stefania.zanet@unito.it

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This article is distributed under the terms of the Creative Commons Attribution Noncommercial License (CC BY-NC 4.0) which permits any noncommercial use, distribution, and reproduction in any medium, provided the original author(s) and source are credited. lates species was also developed. The overall area suitable for *Babesia* spp. in this simulation was found to be 3,723 km², which corresponds to 15.51% of the background regional territory. The model developed was empirically validated assessing tick abundance in randomly chosen areas classified by the model as moderately or highly suitable for *Babesia* spp. Collected Ixodidae ticks analysed for *Babesia* infection by molecular analysis, to confirm the model predictions, indicated a significantly higher prevalence of this infection in areas identified by the model as highly suitable compared to areas classified as only moderately so $(X^2 = 5.05, p<0.05, Odds Ratio= 2.12, Confidence Interval 95% = 1.1-4.1).$

Introduction

Piroplasmosis, caused by protozoa of the genus Babesia, is gaining increasing attention globally as an emerging zoonosis (Hong et al., 2019; Svensson et al., 2019). Babesia is the most frequent parasite transmitted by blood transfusion in North America (Vannier and Krause, 2012; Ryan et al., 2020) and it causes high morbidity and mortality in livestock in endemic tropical and subtropical areas (Kivaria et al., 2007; Pfeffer et al., 2018). Despite the economic and public health importance, little information is available regarding the parasite-host-vector-environment interface (Pérez De León et al., 2010). The epidemiology of a disease may be complex, especially if a vector is required (e.g., Ixodidae ticks for Babesia spp.) and if many species can be reservoirs or deadend hosts. To better understand the epidemiology of circulating tick-borne piroplasms, we modelled Babesia spp. occurrence in wild ruminants that are among the main hosts of Ixodes ricinus and also play the role of a viable platform for systemic and nonsystemic pathogens transmission among co-feeding ticks (Kiffner et al., 2010). We targeted three host species: roe deer (Capreolus capreolus), red deer (Cervus elaphus) and alpine chamois (Rupicapra rupicapra) for which prevalence data of Babesia spp. infection, within the study area, were available (Zanet et al., 2014). C. capreolus is the host species with the highest rates of infection, ranging from 12.6% to 83.6% (Bonnet et al., 2007; Tampieri et al., 2008; Malandrin et al., 2010; Zanet et al., 2014; Razanske et al., 2019). It is the reservoir host of B. capreoli (Yabsley and Shock, 2013) and of the zoonotic B. venatoroum (Malandrin et al., 2010). Red deer and chamois are highly susceptible to Babesia infection (Yabsley and Shock, 2013). In alpine chamois frequent fatal cases have been reported (Hoby et al., 2007, 2009) as consequences of spillover from sympatric roe deer (Hoby et al., 2007, 2009; Tampieri et al., 2008). In north-western Italy (the Piedmont region) the roe deer population is experiencing a steady increase and territorial expansion (Carnevali et al., 2009) that have driven a concomitant increase of ticks, especially I. ricinus which is the most common tick of temperate regions (Vor et al., 2010; Rizzoli et al., 2014).







The modelling effort of the present study was focused on the environmental requirements of this tick species (Estrada-Peña, 2001a). Species Distribution Models (SDM) are produced by relating species occurrence data (i.e. Babesia infection) to a set of environmental variables reflecting key factors for species occurrence, such as climate, topography, geology or land-cover. Ecological Niche Models (ENMs) describe the niche of a species by describing its most suitable habitat (Pulliam, 2000; Sillero, 2011). We developed separate models for cervids (roe deer and red deer) and chamois to account for the different roles of reservoir and spillover hosts, respectively. The proposed ENM for Babesia spp. can serve as tool in human and veterinary health to understand the distribution of Babesia spp. presence, to improve diagnosis of cryptic clinical cases in humans and to facilitate the efficacy of monitoring, control and preventive efforts. The developed ENM was experimentally evaluated by assessing tick abundance in the environment and by analysing ticks collected in areas identified by the model as moderately or highly suitable for Babesia spp.

Materials and Methods

Study site

Species distribution modelling

Table 1. Candidate modelling predictors.

Factors affecting the distribution of Babesia spp. were estimated using MaxEnt 3.3.3 (Phillips et al., 2006). The algorithm implemented in MaxEnt estimates a target probability distribution (i.e. Babesia spp.) by finding the probability distribution of maximum entropy (i.e. most spread out or closest to uniform) within a set of constraints that represent the available information about the environmental requirements of our target distribution (Phillips et al., 2006). We adhered to the default settings for the regularization multiplier, maximum number of iterations, convergence threshold and maximum number of background points that were limited to the Piedmont administrative division with exclusion of all urban areas as unsuitable for piroplasms hosts and vectors (Hirzel et al., 2006; Elith et al., 2011). We generated models randomly assigning 75% of occurrences as training data with the remaining 25% used as test data. We ran five cross-validate replicates for each model. Selection of 'features' (predictors) was carried out automatically following the default rules dependent on the number of presence records (Phillips et al., 2006; Elith et al., 2011). We used the logistic output format conditioned on the environmental variables in each grid cell with suitability values ranging from 0 (unsuitable

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environmental condition) to 1 (optimal environmental condition) (Hu and Jiang, 2011). The threshold value above which there is a substantial probability of presence of *Babesia* spp. was set for each model to the value that maximizes training sensitivity and specificity (Phillips *et al.*, 2006; Elith *et al.*, 2011). Values of habitat suitability above the threshold were divided in four quartiles for graphical visualization and further on-field validation. Model evaluation and best-model selection were carried out using the Area Under the receiver-operator Curve (AUC) (Merow *et al.*, 2013) and lower model complexity (Elith and Leathwick, 2009). The best performing model (Δ AUC \leq 2) and lower number of covariates was chosen. Covariates to be retained were selected by backward stepwise model selection excluding the feature with the lowest permutation importance at each step (Merow *et al.*, 2013).

Infection: presence only data

A total of 61 animals, positive by the polymerase chain reaction (PCR) for Babesia spp., were used as Presence Points (PP) to train the MaxEnt model. Specifically, roe deer (n=30), red deer (n=23) and alpine chamois (n=8) are the three ungulate species included in the model. Positivity to Babesia spp. in these animals has been assessed previously (Zanet et al., 2014). The PP were georeferenced by Global Positioning System (GPS) coordinates (UTM, datum ED50, Fuse 32N) which refer to the location where the animals were killed by hunting or found dead. We developed three models to estimate the probability of presence of Babesia spp. in Piedmont. The first one considered only cervids (Model 1) as target hosts, and we used PP data of Babesia spp. isolated from either red deer or roe deer. Presence of red deer and roe deer habitat can overlap, whereby tick-borne diseases are shared among both species (Duh et al., 2005; Tampieri et al., 2008; Bastian et al., 2012). Cervids are the primary source of blood meals for nymphs and adult I. ricinus (Cadenas et al., 2007; Humair et al., 2007) and their abundance is directly related to deer presence and abundance in the area (Wilson et al., 1990). The second model was trained with alpine chamois PP data only, as this species is usually found at higher elevations and in an environment different from where cervids are typically found, even if a partial overlap between the two species can occur (Model 2). For *Babesia* spp, chamois is primarily a dead-end, spillover host. (Hoby et al., 2009). The third model used all PP records from cervids and chamois together to summarize Babesia spp. circulation in the three most frequently infected wild ungulate species in Piedmont (Model 3).

Candidate predictors

A total of 21 predictors were selected based on the ecological requirements of *Babesia* definitive hosts and vectors (Table 1).

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Host species predictor	Tick vector predictor
Roe deer hunting bag consistency	Altitude
Red deer hunting bag consistency	Land cover
Chamois hunting bag consistency	Annual solar exposure
Ungulate hunting bag consistency	Summer/winter brightness temperature (min, max, mean)
	Summer/winter NDVI (min, max, mean)
	Surface aspect
	Surface slope

A total of 21 host (n=4) and environmental (n=17) factors were evaluated as potential predictors for MaxEnt models



Biotic and abiotic factors influencing the presence of *I. ricinus* were chosen considering the available information on piroplasmid infections in wild ungulates in Europe (Duh *et al.*, 2005; Hoby *et al.*, 2007, 2009; Tampieri *et al.*, 2008; Malandrin *et al.*, 2010; Zintl *et al.*, 2011; Bastian *et al.*, 2012; Overzier *et al.*, 2013; Rizzoli *et al.*, 2014; Ebani *et al.*, 2016; Cézanne *et al.*, 2017; Kauffmann *et al.*, 2017; Remesar *et al.*, 2019). The spatial resolution of all predictors as well as of that of the final model was 1000 m.

Host species predictors

Presence and hunting bag numerical consistency (roe deer, red deer and chamois) were inferred from the Official Hunting dataset (Osservatorio Faunistico Regione Piemonte).

Tick vector predictors

Five abiotic candidate priors were included in the model: i) altitude (ALT); ii) annual solar exposure (total KWatthour/m² received during a year period (SOLAR); iii) surface aspect (slope direction of the maximum rate of change in value from each cell to its neighbours (ASPECT); iv) slope (steepness of land surface measured in degrees from the horizontal (SLOPE) - original resolution 250 m (http://webgis.arpa.piemonte.it); and v) Brightness Temperature (BT) - resolution 1000 m (http://glovis.usgs.gov). Among the biotic priors, the performance of the Normalized Difference Vegetation Index (NDVI) and land cover were tested. NDVI was considered separately for two periods: i) from May to September (vegetation peak season, n=34 Landsat images) and ii) from November to March (minimum vegetation presence, n=29 Landsat images) - resolution 1000 m (http://glovis.usgs.gov). For each season, we evaluated the fit of PP data on the minimum, maximum and mean NDVI that was used as indicator of vegetation coverage and soil relative humidity (Estrada-Peña, 2001a). The same Landsat image collection was used for calculating summer and winter BT (minimum, maximum and mean BT) as indicator of the relative ground temperature (Eisen *et al.*, 2010; Hönig *et al.*, 2011). The land cover raster map (http://www.ruparpiemonte.it/ geocatalogorp/main/?sezione=catalogo) was also evaluated as a model covariate (LAND COVER). To exclude autocorrelation among covariates (cor), independence was assessed using a correlation matrix.

Output analysis and visualization

MaxEnt graphical outputs were analyzed and visualized using QuantumGIS 3.4 (QGIS Developmental Team, 2018). The statistical analysis was performed using R 3.4.4 (R Development Core Team, 2018).

Experimental model evaluation of questing ticks

Tick collection and biomolecular analysis

To experimentally confirm the validity of the model, we randomly selected 8 locations uniformly distributed across the study area which were classified by Model 3 as moderately suitable (locations Q2 - 4) and highly so (locations Q4 - 4) for *Babesia* spp. Model 3 was used as reference for empirical validation to include the broadest possible range for *Babesia* spp. circulation within the study area. In each of these areas, we sampled monthly for questing Ixodidae ticks by dragging. Forty transects (5 transects for each location, each transect covering an area of 100 m²) were sampled each month from May to October. Before each sampling session the exact coordinates of the sampled transect, temperature and humidity were recorded. Collected ticks were preserved in 70% ethanol and identified with reference to species, life-stage and gender under a stereomicroscope using appropriate morphological keys (Walker, 2003; Estrada-Peña et al., 2017). Identified ticks were divided into pools comprising specimens collected from the same transect and homogeneous for species, developmental stage and sex. The tick pools were mechanically homogenized using Qiagen TissueLyser LT (Qiagen, Milan, Italy). Total genomic DNA was extracted according to Maurelli et al. (2018) and the V4 hypervariable region of the 18S rDNA amplified by a semi-nested PCR methodology (Zanet et al., 2017). All PCR-positive amplicons were purified using a commercial kit (Nucleospin Extract II Kit, Macherey-Nagel, Düren, Germany) and both strands sequenced (Macrogen Europe, Spain) for species identification. The resulting nucleotide sequences were analyzed using MEGA X software (Kumar et al., 2018) and compared to those available in GenBank (http://www.ncbi.nlm.nih.gov/genbank). The PCR results were expressed as a minimum infection rate (MIR) or the minimum percentage of ticks in a pool with detectable DNA of Babesia spp. (Kramer et al., 1999).

Statistical analysis

Mean tick abundance in the Q2 and Q4 locations was compared using Student's T test for paired data. To identify variables associated with *Babesia* spp. infection in ticks collected in the environment we calculated using Chi Square test (X^2) and Odd Ratio (OR). Differences were considered significant at p<0.05. All analysis was performed using R 3.5.3 (R Development Core Team, 2018).

Results

Assessment of covariate independence

The correlation matrix of covariates dependence allowed us to exclude from contemporary use in MaxEnt analysis SLOPE and ALT (cor=0.71), maximum summer BT and ALT (cor=0.70).

Model 1 - Babesia spp. distribution in cervids

The best performing model (AUC= 0.937, Figure 1) for Babesia spp. occurrence in cervids was generated using the following parameters as covariates: roe deer and red deer abundance, mean summer NDVI, SLOPE, SOLAR and LAND COVER (see Figure 2 for the response curves of each feature and for the jackknife test of variable importance on training gain and on AUC). Red deer abundance, together with mean summer NDVI, SLOPE and SOLAR had a direct correlation with Babesia spp. occurrence probability. Babesia spp. suitability peaked with roe deer abundance of 4 individuals/km² and the most suitable land use categories were found to be broad-leaved forest (category 9), ecotonal shrub areas (category 10), high altitude pastures and meadows (category 14) and grasslands (category 7). The suitable area for Babesia spp. infection in wild cervids was estimated at 956 km² (about 4% of the regional territory; threshold value for Maximum Training Sensitivity plus Specificity 0.443) (Table 2, Figure 3).

Model 2- Babesia spp. distribution in alpine chamois

The best performing model (AUC= 0.993, Figure 1) for *Babesia* spp. in Alpine chamois was generated using the following parameters as covariates: chamois density, ALT, SOLAR LAND





COVER and mean summer NDVI (see Figure 3 for the response curves of each feature, and the jackknife test of variable importance on training gain and on AUC). All continuous variables had a direct correlation with *Babesia* spp. probability, while the land cover categories most suitable to infection were found to be high altitude pastures and meadows (category 14), grasslands (category 7) and ecotonal shrub areas (category 10). The suitable area for *Babesia* spp. infection in chamois was estimated at 81 km² (0.35% of the regional territory; threshold value for Maximum Training Sensitivity plus Specificity 0.465) (Table 3).

Table 2. Suitable area for Babesia spp. infection in cervids.

Probability of presence	Area (km ²)	Percent of the area of presence	Percent of the area of presence
1 st Quartile	206	21.58%	0.86%
2 nd Quartile	211	22.09%	0.88%
3 rd Quartile	226	23.65%	0.94%
4 th Quartile	313	32.68%	1.30%
Total area of presence	956	100.00%	3.98%

The predicted presence area for Babesia spp. infection was divided into quartiles of growing probability of presence. The size of the suitable area (km²), the percent of the total presence area (above minimum threshold of 0.443) and the percent on the background (Piedmont Region without urbanized areas) is given for each quartile.

Table 3. Suitable area for Babesia spp. infection in chamois.

Probability of presence	Area (km ²)	Percent of the area of presence	Percent of the area of presence		
1 st Quartile	29	35.19%	0.12%		
2 nd Quartile	18	22.30%	0.08%		
3 rd Quartile	17	20.37%	0.07%		
4 th Quartile	17	22.15%	0.08%		
Total area of presence	81	100.00%	0.35%		

The predicted presence area for Babesia spp. infection was divided into quartiles of growing probability of presence. The size of the suitable area (km²), the percent of the total presence area (above minimum threshold of 0.443) and the percent on the background (Piedmont Region without urbanized areas) is given for each quartile.







Figure 1. ROC analysis and AUC values. Performance of each of the three models of *Babesia* spp. probability of infection in a. cervids; b. chamois; c. all ungulate species) evaluated by receiver operating characteristic (ROC) analysis, which characterizes the performance of a model at all possible thresholds by a single AUC value (Area Under the Curve).





Model 3 - Overall Babesia spp. distribution in Piedmont

The occurrence of *Babesia* spp. was also estimated for cervids and chamois together, as there is evidence that natural infection in alpine chamois is frequently acquired from sympatric roe deer (Hoby *et al.*, 2007, 2009; Tampieri *et al.*, 2008; Zanet *et al.*, 2014). The overall suitable area for *Babesia* spp. in this simulation was estimated at 3,723 km² which corresponds to approximately 15.5% of the background regional territory (Figure 4). The AUC of the model was 0.91 (Figure 1). distribution of occurrence probability is reported in Table 4. Retained covariates were: mean summer NDVI, SOLAR, and LAND COVER (Figure 5). The probability of presence for *Babesia* spp., peaked when NDVI reaches 0.65, value that corresponds to broad-leaved forests areas, that are notoriously the most suitable environments for *Ixodidae* ticks(Daniel *et al.*, 1998; Estrada-Peña, 2001b; Estrada-Peña and Venzal, 2006). Land cover confirmed the high suitability of broad-leaved forest areas (class 9, in the LAND COVER predictor) and solar exposure showed a direct correlation with increasing *Babesia* spp. habitat suitability.

Field model validation

Over the 6 months of sampling, we collected 1,555 ticks, which were morphologically identified as *I. ricinus* (n=1,553) and

Table 4. Suitable area for Babesia spp. infection in wild ungulates.

Probability of presence	Area (km ²)	Percent of the area of presence	Percent of the area of presence
1 st Quartile	883	23.70%	3.68%
2 nd Quartile	824	22.13%	3.43%
3 rd Quartile	899	24.16%	3.75%
4 th Quartile	1,117	30.01%	4.65%
Total area of presence	3,723	100.00%	15.51%

The predicted presence area for Babesia spp. infection was divided into quartiles of growing probability of presence. The size of the suitable area (km²), the percent of the total presence area (above minimum threshold of 0.443) and the percent on the background (Piedmont Region without urbanized areas) is given for each quartile.



Responce curves of features for *Babesia* sp. probability of presence in CERVIDS



Figure 2. Covariate response curves and Jackknife test of variable importance – Model 1 cervids. Response curves of the selected covariates for *Babesia* spp. occurrence in cervids. Environmental covariates (mean summer NDVI, solar exposure and slope) and host abundance are directly related to increased *Babesia* spp. suitability. Land cover classes mostly associated with Babesia spp. suitability are pasture (cat.7), broad-leaved forest (cat.9), shrubs (cat. 10), and alpine meadows and pastures (cat.14). Jackknife test of variable importance shows the impact of each variable on training gain and AUC. Values shown are averages over replicate runs.







I. hexagonus (n=2). The number of ticks collected from each location is reported in Table 5. Larvae were the most abundant lifestage in all sampling locations (n=1302), followed by nymphs (n=250) and adults (n=3). Identified ticks were divided, for molecular analysis, in 161 homogeneous pools (pools of adult ticks n=3, nymphs n=74 and larvae n=84). Babesia spp. DNA was detected in 59 pools (MIR=36.65%; 95% CI 29.6-44.32). The highest MIR (p<0.05) was recorded in adults (MIR=100.00%; 95% CI 43.85-100) followed by larvae (MIR=36.90%; 95% CI 27.37-47.58) and nymphs (MIR=33.78%; 95% CI 24.05-45.12). From locations moderately suitable for Babesia spp. (Q2) we consistently recovered a lower number of ticks compared to highly suitable areas (Q4) with exception of one Q4 location (location F, Table 5) where no ticks were recovered. Results from location F were possibly impaired by the use of part of the sampling area as pasture for cattle. Babesia spp. MIR was significantly higher in Q4 areas than in Q2 areas (X^2 =5.05 p<0.05) with OR=2.12 (95% CI 1.10-4.10). The zoonotic *B. venatorum* was the most prevalent species with a MIR of 23.60% (95% CI 17.71 – 30.73). *B. capreoli* was reported with an overall MIR of 3.11% (95% CI 1.33 – 7.06) followed by *B. microti* and *B. vulpes*, which were both detected with a MIR of 2.48% (95% CI 0.97 – 6.21%). Protozoa of the genus *Theileria (T. buffeli* group) were detected in 8 pooled samples with a MIR of 4.97% (95% CI 2.54 – 9.50%). All species were detected with a higher prevalence in Q4 areas compared to Q2 (Table 6).

Discussion

The models presented are intended as a tool to better understand the geographical occurrence and epidemiology of *Babesia* piroplasms infecting wild ungulates. Babesiosis plays a key pathogenic role in livestock (Criado-Fornelio *et al.*, 2003; Schnittger *et al.*, 2003; Bock *et al.*, 2008) and as emerging disease









Figure 3. Covariate response curves and Jackknife test of variable importance – Model 2 chamois. Response curves of the selected covariates for *Babesia* spp. occurrence in chamois. Environmental covariates (mean summer NDVI, solar exposure and elevation) and host abundance (chamois) are directly related to higher *Babesia* spp. suitability. Land Cover classes mostly associated to *Babesia* spp. suitability are pastures and Alpine meadows (cat.7 and 14), and shrubs (cat. 10). Jackknife test of variable importance show the impact of each variable on training gain and AUC. Values shown are averages over replicate runs. The environmental variable with highest gain when used in isolation is Land Cover, which therefore appears to have the most useful information by itself, and it is the environmental variables.





in humans (Ryan *et al.*, 2020). *Babesia* infections are still very much unknown, especially with respect to epidemiology in wildlife hosts and wildlife-related ticks and environments (Yabsley and Shock, 2013). The model covers a variety of different environments: from low-altitude broadleaved forest and agricultural areas, to high-altitude Alpine areas. The resolution of the model was high (1000 m) in order to correctly describe and predict suitable areas for highly selective tick vectors which can persist only in particular areas where a set of environmental characteristics has to occur together with the presence of suitable animal hosts. Biotic and abiotic priors were selected and modelled with regard to the requirements of *I. ricinus*, which is associated with rural environments and wild ungulates (Rizzoli *et al.*, 2014). *I. ricinus* was also the species that was most often recovered (P=99.87%, 95% CI 99.53 – 99.96) from the study region.

All three models fitted the variance of our data well (AUC values ≥ 0.91). All the covariates included in the models are biologi-

cally sound and highly responsive in depicting suitable presence areas. NDVI (with the mean value computed during summer months), slope, solar exposure, altitude and land cover were the most informative environmental features. Each of the three models was best described by different priors, although the response of all variables in each model indicates how areas more exposed to sun light, with a rich vegetation coverage (shrub and broad-leaf forest areas), are the most suitable areas for Babesia infections to occur. Pastures and alpine meadows were also found to be suitable for Babesia spp., but this is due to the presence in the model of alpine chamois, whose typical home range includes high-altitude open areas. The suitable area for Babesia to occur is overall extensive and corresponds to 15% of the territory considered in model development. Human infection occurs mostly in rural or natural environments due to the changes in social and economic behaviours (increase of outdoor activities and re-naturalization of rural areas and territorial expansion of wildlife) (Randolph, 2010; Semenza et



Figure 4. *Babesia* spp. Habitat Suitability Model and empiric validation sampling points. Predicted geographical distribution of the suitable areas for *Babesia* spp. in the three ungulate species. The unsuitable area (below the threshold value of minimum probability of presence) in white, while the suitable area (above the threshold) is divided into quartiles of growing suitability from 1st Q- minimum (light green) to 4th Q- maximum suitability (blue). Tick sampling areas shown in insert (Q2 - from A to D and Q4 - from E to F).







Table 5. Ticks and Babesia spp. infection of field model validation.

Site Gepgraphic identification (coordinates)	Number of larvae [no. of pools	MIR Babesia (95% Cl)] [no. of positive pools]	Number of nymphs [no. of positive pools]		Number of adults [no. of positive pools]	MIR Babesia (95% CI) [no. of positive pools]	Total MIR (95% CI) [total no. of positive pools]
Q2	224 [38]	26.32% (14.97-42.01) [10]	8 [37]	27.03% (15.4-42.98) [10]	1 [1]	100%(20.65-100) [1]	27.63% (18.84-38.58) [21/76]
A 45°01'03.9"N 6°48'44.8"E	0 [0]	-	2 [2]	0.00% (0.00-65.76) [0]	0 [0]	-	
B 44°28'50.0"N 7°20'47.5"E	22 [8]	0.00% (0.00-32.44) [0]	27 [11]	18.18% (0.51-47.7) [2]	0 [0]	-	
C 45°04'56.2"N 7°20'35.0"E	34 [9]	44.44% (18.87-73.33) [4]	43 [13]	46.15% (23.21-70.86) [6]	1 [1]	100% (20.65-100) [1]	
D 44°31'45.0"N 8°25'26.7"E	168 [21]	28.57% (13.81-49.96) [6]	15 [11]	18.18% (5.14-47.7) [2]	0 [0]	-	
Q4	1,078 [46]	45.65% (32.15-59.82) [21]	163 [37]	40.54% (26.35-56.51) 15]	2 [2]	100% (34.24-100) [2]	44.71% (34.59-55.28) [38/85]
E 45°06'13.3"N 6°55'04.3"E	284 [19]	31.58% (15.36-53.99) [6]	54 [13]	15.38% (4.33-42.23) [2]	1 [1]	100% (20.65-100) [1]	
F 44°29'17.8"N 7°02'22.4"E	0 [0]	-	0 [0]	-	0 [0]	-	
G 45°01'46.7"N 7°17'12.7"E	238 [15]	73.33% (40.05-89.10) [11]	70 [16]	68.75% (44.4-85.84) [11]		100% (20.65-100) [1]	
H 44°31'02.4"N 8°26'53.3"E	556 [12]	33.33% (13.81-60.94) [4]	39 [8]	25% (7.15-59.07) [2]	0 [0]	-	
Total	1,302 [84]	36.90% (27.37-47.58) [31]	250 [74]	33.78% (24.05-45.12) [25]	3 [3]	100% (43.85-100) [3]	36.65% (29.6-44.32) [59/161]

For each geographically marked location (A to H) identified by Model 1 as moderately suitable (Q2) or highly suitable (Q4), the number of larvae, nymphs and adults of Ixodidae ticks collected and the number of homogeneous pools are reported. MIR = minimum infection rate (%); CI = confidence intervals.

Table 6. Babesia species identification.

Total no. of positive pools (MIR) [95% CI]	Number of positive pools at Q2 (MIR) [95% CI]	Number of positive pools at Q4 (MIR) [95% CI]			
38 (23.60%) [17.71-30.73]	14 (18.42%) [11.30-28.58]	24 (28.24%) [19.77-38.58]	100%	99.5 %	KX857480
4 (2.48%) [0.97-6.21]	1 (1.32%) [0.23-7.08]	3 (3.53%) [1.21-9.87]	100%	100%	FJ608739
4 (2.48%) [0.97-6.21]	1 (1.32%) [0.23-7.08]	3 (3.53%) [1.21-9.87]	100%	100%	KI175166
5 (3.11%) [1.33-7.06]	2 (2.63%) [0.72-9.10]	3 (3.53%) [1.21-9.87]	100%	100%	KU145465
8 (4.97%) [2.54-9.50]	3 (3.95%) [1.35-10.97]	5 (5.88%) [2.54-13.04]	100%	100%	MH327771, AJ616717, KX965721
	pools (MIR) [95% CI] 38 (23.60%) [17.71-30.73] 4 (2.48%) [0.97-6.21] 4 (2.48%) [0.97-6.21] 5 (3.11%) [1.33-7.06]	pools (MIR) [95% CI] at Q2 (MIR) [95% CI] 38 (23.60%) [17.71-30.73] 14 (18.42%) [11.30-28.58] 4 (2.48%) [0.97-6.21] 1 (1.32%) [0.23-7.08] 4 (2.48%) [0.97-6.21] 1 (1.32%) [0.23-7.08] 5 (3.11%) [1.33-7.06] 2 (2.63%) [0.72-9.10]	pools (MIR) [95% CI] at Q2 (MIR) [95% CI] at Q4 (MIR) [95% CI] 38 (23.60%) [17.71-30.73] 14 (18.42%) [11.30-28.58] 24 (28.24%) [19.77-38.58] 4 (2.48%) [0.97-6.21] 1 (1.32%) [0.23-7.08] 3 (3.53%) [1.21-9.87] 4 (2.48%) [0.97-6.21] 1 (1.32%) [0.23-7.08] 3 (3.53%) [1.21-9.87] 5 (3.11%) [1.33-7.06] 2 (2.63%) [0.72-9.10] 3 (3.53%) [1.21-9.87]	pools (MIR) [95% CI] at Q2 (MIR) [95% CI] at Q4 (MIR) [95% CI] coverage 38 (23.60%) [17.71-30.73] 14 (18.42%) [11.30-28.58] 24 (28.24%) [19.77-38.58] 100% 4 (2.48%) [0.97-6.21] 1 (1.32%) [0.23-7.08] 3 (3.53%) [1.21-9.87] 100% 4 (2.48%) [0.97-6.21] 1 (1.32%) [0.23-7.08] 3 (3.53%) [1.21-9.87] 100% 5 (3.11%) [1.33-7.06] 2 (2.63%) [0.72-9.10] 3 (3.53%) [1.21-9.87] 100%	pools (MIR) [95% CI] at Q2 (MIR) [95% CI] at Q4 (MIR) [95% CI] coverage identity 38 (23.60%) [17.71-30.73] 14 (18.42%) [11.30-28.58] 24 (28.24%) [19.77-38.58] 100% 99.5 % 4 (2.48%) [0.97-6.21] 1 (1.32%) [0.23-7.08] 3 (3.53%) [1.21-9.87] 100% 100% 4 (2.48%) [0.97-6.21] 1 (1.32%) [0.23-7.08] 3 (3.53%) [1.21-9.87] 100% 100% 5 (3.11%) [1.33-7.06] 2 (2.63%) [0.72-9.10] 3 (3.53%) [1.21-9.87] 100% 100%

Number of homogeneous tick pools positive for each species. MIR = minimum infection rate (%); CI = confidence intervals.



Responce curves of features for Babesia sp. propability of presence in UNGULATES

Jackknife test of variables importance



Figure 5. Covariate response curves and Jackknife test of variable importance – Model 3 ungulates. Response curves of the selected covariates for overall *Babesia* spp. occurrence in the three studies species of ungulates. Environmental covariates (mean summer NDVI and solar exposure) are directly related to increased Babesia spp. suitability, while Land Cover classes mostly associated with *Babesia* spp. suitability are pasture (cat.7), broad-leaved forest (cat.9), shrubs (cat. 10), and alpine meadows and pastures (cat.14). Jackknife test of variable importance show the impact of each variable on training gain and AUC. Values shown are averages over replicate runs.





al., 2016). In the last decades, the population of wild ungulates, especially that of roe deer in Italy, has been growing at increasing rates (26% between the years 2000 and 2005; Carnevali *et al.*, 2009). Together with ungulates, the number of ticks has also increased and their geographical range has expanded to include areas at higher latitude and/or altitudes, possibly due to the influence of climate changes on tick habitats (Tälleklint and Jaenson, 1998; Lindgren *et al.*, 2000; Daniel *et al.*, 2003; Estrada-Peña and Venzal, 2006; Materna *et al.*, 2008; Jore *et al.*, 2014). A study by Stainforth *et al.* (2013) documents a decrease between 5% and 10% of the number of winter nights below zero in the study area. Higher winter temperatures favour overwintering and extend the activity season of the vector ticks (Materna *et al.*, 2008) thus increasing the risk of tick-borne diseases.

Field sampling of Ixodidae ticks in the study area was used to validate the Babesia spp. occurrence model with empirical data. Results highlighted a significant positive association (p < 0.05) between Babesia spp. MIR in collected ticks and highly suitable areas for parasite occurrence (Q4 areas), thus confirming the ability of the model to correctly predict habitat-suitability for Babesia spp. The higher prevalence of infection in the Q4 area than the Q2 areas was confirmed for each of the piroplasmid species identified by sequencing. Most of the species of Babesia identified in sampled ticks had been previously reported for the study area. The roe deer used as PP, were infected with B. venatorum, B. capreoli and B. bigemina (Zanet et al., 2014). Roe deer is the main reservoir host for both B. venatorum and B. capreoli (Malandrin et al., 2010; Michel et al., 2014), while B. bigemina, together with B. divergens, is mainly associated with bovine babesiosis (Zintl et al., 2003; Hilpertshauser et al., 2007). B. capreoli was the most prevalent species found in red deer used as PP followed by Theileria spp., as has been shown previously (Zanet et al., 2014); this species was also detected in the ticks used for empirical validation. All chamois used as PP were infected with B. capreoli, as has also been shown previously (Zanet et al., 2014). B. vulpes has been reported to infect Red foxes and horses in the same study area (Zanet et al., 2014, 2017). This species is also recognized as a pathogenic in dogs resulting in symptoms of clinical infection ranging from pale mucous membranes, anorexia, apathy to fever with severe macrocytic/hypochromic regenerative anaemia and thrombocytopenia (Guitián et al., 2003; Baneth et al., 2019). Two zoonotic piroplasm species were detected in ticks, namely B. venatorum and B. microti. I. ricinus is the main vector for both these species (Gray et al., 2002; Bonnet et al., 2007). B. venatorum was detected in 6 out of the 8 sampling locations, both in Q2 and Q4 areas. It was not detected only at sites B (Q2) and F (Q4) where Babesia spp. were generally detected with lower MIR values or no ticks collected at all. B. microti was instead reported only at locations C (Q2) and G (Q4) where Babesia spp. with the highest MIR in larvae and nymphs pools were also found.

Conclusions

The use of ecological niche modelling for infectious diseases is a useful tool that allows identifying and prioritizing interventions for disease surveillance, prevention and control. This kind of modelling is highly recommendable for decision makers, managers and politician as it contributes to better use of limited resources.

References

- Baneth G, Cardoso L, Brilhante-Simões P, Schnittger L, 2019. Establishment of Babesia vulpes n. sp. (Apicomplexa: Babesiidae), a piroplasmid species pathogenic for domestic dogs. Parasite Vector 12:1–8. doi.org/10.1186/s13071-019-3385-z
- Bastian S, Jouglin M, Brisseau N, Malandrin L, Klegou G, L'Hostis M, Chauvin A, 2012. Antibody prevalence and molecular identification of Babesia spp. in roe deer in France. J Wildl Dis 48:416–24. doi.org/10.7589/0090-3558-48.2.416
- Bock RE, Jackson LA, de Vos AJ, Jorgensen WK, 2008. Babesiosis of cattle, in: Ticks: Biology, Disease and Control. doi.org/10.1017/CBO9780511551802.014
- Bonnet S, Jouglin M, L'Hostis M, Chauvin A, 2007. Babesia sp. EU1 from roe deer and transmission within Ixodes ricinus. Emerg Infect Dis 13(8):1208. doi.org/10.3201/eid1308.061560
- Cadenas FM, Rais O, Humair PF, Douet V, Moret J, Gern L, 2007. Identification of host bloodmeal source and Borrelia burgdorferi sensu lato in field-collected Ixodes ricinus ticks in Chaumont (Switzerland). J Med Entomol 44, 1109–1117. doi.org/10.1093/jmedent/44.6.1109
- Carnevali L, Pedrotti L, Riga F, Toso S, 2009. Banca Dati Ungulati. Status, distribuzione, consistenza, gestione e prelievo venatorio delle popolazioni di ungulati in Italia. Rapporto 2001-2005 [Ungulates in Italy. Status, distribution, abundance, management and hunting of ungulate populations in Italy]. Biol Cons Fauna 117, 1-168.
- Cézanne R, Mrowietz N, Eigner B, Duscher GG, Glawischnig W, Fuehrer HP, 2017. Molecular analysis of Anaplasma phagocytophilum and Babesia divergens in red deer (Cervus elaphus) in Western Austria. Mol Cell Probes 31, 55–58. doi.org/ 10.1016/j.mcp.2016.07.003
- Criado-Fornelio A, Martinez-Marcos A, Buling-Saraña A, Barba-Carretero JC, 2003. Molecular studies on Babesia, Theileria and Hepatozoon in southern Europe: Part I. Epizootiological aspects. Vet Parasitol 113, 189–201. doi.org/10.1016/S0304-4017(03)00078-5
- Daniel M, Danielová V, Kříž B, Jirsa A, Nožička J, 2003. Shift of the tick Ixodes ricinus and tick-borne encephalitis to higher altitudes in Central Europe. Eur J Clin Microbiol Infect Dis 22, 327–328. doi.org/10.1007/s10096-003-0918-2
- Daniel M, Kol J, Zeman P, Pavelka K, Sadlo J, 1998. Predictive map of Ixodes ricinus high-incidence habitats and a tick-borne encephalitis risk assessment using satellite data. Exp Appl Acarol 22, 417–433. doi.org/10.1023/a:1006030827216
- Duh D, Petrovec M, Bidovec A, Avsic-Zupane T, 2005. Cervids as Babesiae hosts, Slovenia. Emerg Infect Dis 11, 1121–1123. doi.org/10.3201/eid1107.040724
- Ebani VV, Rocchigiani G, Bertelloni F, Nardoni S, Leoni A, Nicoloso S, Mancianti F, 2016. Molecular survey on the presence of zoonotic arthropod-borne pathogens in wild red deer (Cervus elaphus). Comp Immunol Microbiol Infect Dis 47, 77–80. doi.org/10.1016/j.cimid.2016.06.003
- Eisen RJ, Eisen L, Girard YA, Fedorova N, Mun J, Slikas B, Leonhard S, Kitron U, Lane RS, 2010. A spatially-explicit model of acarological risk of exposure to Borrelia burgdorferiinfected Ixodes pacificus nymphs in northwestern California based on woodland type, temperature, and water vapor. Ticks Tick-borne Dis 1, 35–43. doi.org/10.1016/j.ttbdis.2009.12.002 Elith J, Leathwick JR, 2009. Species Distribution Models: ecolog-







ical explanation and prediction across space and time. Annu Rev Ecol Evol Syst 40, 677–697. doi.org/10.1146/annurev. ecolsys.110308.120159

- Elith J, Phillips SJ, Hastie T, Dudík M, Chee YE, Yates CJ, 2011. A statistical explanation of MaxEnt for ecologists. Divers Distrib 17, 43–57. doi.org/10.1111/j.1472-4642.2010.00725.x
- Estrada-Peña A, 2001a. Forecasting habitat suitability for ticks and prevention of tick-borne diseases. Vet Parasitol 98, 111–132. doi.org/10.1016/S0304-4017(01)00426-5
- Estrada-Peña A, 2001b. Distribution, abundance and habitat preferences of Ixodes ricinus (Acari: Ixodidae) in Northern Spain. J Med Entomol 38, 361–370. doi.org/10.1603/0022-2585-38.3.361
- Estrada-Peña A, Venzal JM, 2006. Changes in habitat suitability for the tick Ixodes ricinus (Acari: Ixodidae) in Europe (1900-1999). Ecohealth 3, 154–162. doi.org/10.1007/s10393-006-0036-6
- Estrada-Peña A, Mihalca AD, Petney T, 2017. Ticks of Europe and North Africa: A guide to Species Identification. doi.org/10.1007/978-3-319-63760-0
- Gray J, Von Stedingk LV, Gürtelschmid M, Granström M, 2002. Transmission studies of Babesia microti in Ixodes ricinus ticks and gerbils. J Clin Microbiol doi.org/10.1128/JCM.40.4.1259-1263.2002
- Guitián FJ, Camacho AT, Telford SR, 2003. Case-control study of canine infection by a newly recognised Babesia microti-like piroplasm. Prev Vet Med 61, 137–145. doi.org/10.1016/S016 7-5877(03)00164-8
- Hilpertshauser H, Deplazes P, Meli ML, Hofmann-Lehmann R, Lutz H, Mathis A, 2007. Genotyping of Babesia bigemina from cattle from a non-endemic area (Switzerland). Vet Parasitol 145, 59–64. doi.org/10.1016/j.vetpar.2006.12.006
- Hirzel AH, Le Lay G, Helfer V, Randin C, Guisan A, 2006. Evaluating the ability of habitat suitability models to predict species presences. Ecol Modell 199, 142–152. doi.org/10.1016/j.ecolmodel.2006.05.017
- Hoby S, Mathis A, Doherr MG, Robert N, Ryser-Degiorgis MP, 2009. Babesia capreoli infections in alpine chamois (Rupicapra R. Rupicapra), roe Deer (Capreolus C. Capreolus) and red Deer (Cervus elaphus) from Switzerland. J Wildl Dis 45, 748–753. doi.org/10.7589/0090-3558-45.3.748
- Hoby S, Robert N, Mathis A, Schmid N, Meli ML, Hofmann-Lehmann R, Lutz H, Deplazes P, Ryser-Degiorgis MP, 2007.
 Babesiosis in free-ranging chamois (Rupicapra r. rupicapra) from Switzerland. Vet Parasitol 148, 341–345. doi.org/10.1016/j.vetpar.2007.06.035
- Hong SH, Kim SY, Song BG, Rho JY, Cho CR, Kim CN, Um TH, Kwak YG, Cho SH, Lee SE, 2019. Detection and characterization of an emerging type of Babesia sp. similar to Babesia motasi for the first case of human babesiosis and ticks in Korea. Emerg Microbes Infect 8, 869–878. doi.org/10.1080/22221751.2019.1622997
- Hönig V, Vec PŠ, Masař O, Grubhoffer L, 2011. Tick-borne diseases risk model for south Bohemia (Czech Republic). GIS Ostrava 2011 1, 23–26. Available at: gis.vsb.cz/gis2011/
- Hu J, Jiang Z, 2011. Climate change hastens the conservation urgency of an endangered ungulate. PLoS One 6(8), 322873. doi.org/10.1371/journal.pone.0022873
- Humair PF, Douet V, Moran Cadenas F, Schouls LM, Van De Pol I, Gern L, 2007. Molecular identification of blood meal source in Ixodes ricinus ticks using 12S rDNA as a genetic marker. J

Med Entomol 44, 869-880.

- Jore S, Vanwambeke SO, Viljugrein H, Isaksen K, Kristoffersen AB, Woldehiwet Z, Johansen B, Brun E, Brun-Hansen H, Westermann S, Larsen IL, Ytrehus B, Hofshagen M, 2014. Climate and environmental change drives Ixodes ricinus geographical expansion at the northern range margin. Parasit Vector 7, 1–14. doi.org/10.1186/1756-3305-7-11
- Kauffmann M, Rehbein S, Hamel D, Lutz W, Heddergott M, Pfister K, Silaghi C, 2017. Anaplasma phagocytophilum and Babesia spp. in roe deer (Capreolus capreolus), fallow deer (Dama dama) and mouflon (Ovis musimon) in Germany. Mol Cell Probes 31, 46–54. doi.org/10.1016/j.mcp.2016.08.008
- Kiffner C, Lödige C, Alings M, Vor T, Rühe F, 2010. Abundance estimation of Ixodes ticks (Acari: Ixodidae) on roe deer (Capreolus capreolus). Exp Appl Acarol 52, 73–84. doi.org/10.1007/s10493-010-9341-4
- Kivaria FM, Ruheta MR, Mkonyi PA, Malamsha PC, 2007. Epidemiological aspects and economic impact of bovine theileriosis (East Coast fever) and its control: a preliminary assessment with special reference to Kibaha district, Tanzania. Vet J 173, 384–390. doi.org/10.1016/j.tvj1.2005.08.013
- Kramer V, Randolph M, Hui L, Irwin W, Gutierrez A, Duc J, 1999.
 Detection of the agents of Human Ehrlichioses in Ixodid ticks from California. Am J Trop Med Hyg 60, 62–65. doi.org/10.4269/ajtmh.1999.60.62
- Kumar S, Stecher G, Li M, Knyaz C, Tamura K, 2018. MEGA X: Molecular Evolutionary Genetics Analysis across computing platforms. Mol Biol Evol 35, 1547–1549. doi.org/10.1093/ molbev/msy096
- Lindgren E, Tälleklint L, Polfeldt T, 2000. Impact of climatic change on the northern latitude limit and population density of the disease-transmitting European tick Ixodes ricinus. Environ Health Perspect 108, 119–123. doi.org/10.1289/ehp.00108119
- Malandrin L, Jouglin M, Sun Y, Brisseau N, Chauvin A, 2010. Redescription of Babesia capreoli (Enigk and Friedhoff, 1962) from roe deer (Capreolus capreolus): isolation, cultivation, host specificity, molecular characterisation and differentiation from Babesia divergens. Int J Parasitol 40, 277–284. doi.org/10.1016/j.ijpara.2009.08.008
- Materna J, Daniel M, Metelka L, Harcarik J, 2008. The vertical distribution, density and the development of the tick Ixodes ricinus in mountain areas influenced by climate changes. Int J Med Microbiol 298, 25–37. doi.org/10.1016/j.ijmm.2008.05.004
- Maurelli MP, Pepe P, Colombo L, Armstrong R, Battisti E, Morgoglione ME, Counturis D, Rinaldi L, Cringoli G, Ferroglio E, Zanet S, 2018. A national survey of Ixodidae ticks on privately owned dogs in Italy. Parasite Vector 11, 420. doi.org/10.1186/s13071-018-2994-2
- Merow C, Smith MJ, Silander JA, 2013. A practical guide to MaxEnt for modeling species' distributions: what it does, and why inputs and settings matter. Ecography 36, 1058–1069. doi.org/10.1111/j.1600-0587.2013.07872.x
- Michel AO, Mathis A, Ryser-Degiorgis MP, 2014. Babesia spp. in European wild ruminant species: parasite diversity and risk factors for infection. Vet Res 45, 1–11. doi.org/10.1186/1297-9716-45-65
- Overzier E, Pfister K, Herb I, Mahling M, Böck G, Silaghi C, 2013. Detection of tick-borne pathogens in roe deer (Capreolus capreolus), in questing ticks (Ixodes ricinus), and in ticks infesting roe deer in southern Germany. Ticks Tick-Borne Dis





4, 320-328. doi.org/10.1016/j.ttbdis.2013.01.004

- Pérez De León AA, Strickman DA, Knowles DP, Fish D, Thacker E, De La Fuente J, Krause PJ, Wikel SK, Miller RS, Wagner GG, Almazn C, Hillman R, Messenger MT, Ugstad PO, Duhaime RA, Teel PD, Ortega-Santos A, Hewitt DG, Bowers EJ, Bent SJ, Cochran MH, McElwain TF, Scoles GA, Suarez CE, Davey R, Howell Freeman JM, Lohmeyer K, Li AY, Guerrero FD, Kammlah DM, Phillips P, Pound JM, 2010. One health approach to identify research needs in bovine and human babesioses: Workshop report. Parasit Vector 3, 1–12. doi.org/10.1186/1756-3305-3-36
- Pfeffer M, Krol N, Obiegala A, 2018. Prevention and control of tick-borne anaplasmosis, cowdriosis and babesiosis in the cattle industry, in: Ecology and Control of Vector-Borne Diseases. Wageningen Academic Pub, 175–194.
- Phillips SJ, Anderson RP, Schapire RE, 2006. Maximum entropy modeling of species geographic distributions. Ecol Modell 190, 231–259. doi.org/10.1016/j.ecolmodel.2005.03.026
- Pulliam HR, 2000. On the relationship between niche and distribution. Ecol Lett 3, 349–361. doi.org/10.1046/j.1461-0248.2000.00143.x
- Randolph SE, 2010. Human activities predominate in determining changing incidence of tick-borne encephalitis in Europe. Euro Surveill 15, 24–31. doi.org/10.2807/ese.15.27.19606-en
- Razanske I, Rosef O, Radzijevskaja J, Bratchikov M, Griciuviene L, Paulauskas A, 2019. Prevalence and co-infection with tickborne Anaplasma phagocytophilum and Babesia spp. in red deer (Cervus elaphus) and roe deer (Capreolus capreolus) in Southern Norway. Int J Parasitol Parasites Wildl.8, 127–134. doi.org/10.1016/j.ijppaw.2019.01.003
- Remesar S, Fernández PD, Venzal JM, Pérez-Creo A, Prieto A, Estrada-Peña A, López CM, Panadero R, Fernández G, Díez-Baños P, Morrondo P, 2019. Tick species diversity and population dynamics of Ixodes ricinus in Galicia (north-western Spain). Ticks Tick- Borne Dis 10, 132–137. doi.org/10.1016/j.ttbdis.2018.09.006
- Rizzoli A, Silaghi C, Obiegala A, Rudolf I, Hubalek Z, Faldvari G, Plantard O, Vayssier-Taussat M, Bonnet S, Åitalska E, Kazimrova M, 2014. Ixodes ricinus and its transmitted pathogens in urban and peri-urban areas in Europe: new Hhzards and relevance for public health. Front Public Health 2, 251. doi.org/10.3389/fpubh.2014.00251
- Ryan ET, Hill DR, Solomon T, Aronson NE, Endy TP, 2020. Hunter's tropical medicine and emerging infectious diseases (10th ed.).
- Schnittger L, Yin H, Gubbels MJ, Beyer D, Niemann S, Jongejan F, Ahmed JS, 2003. Phylogeny of sheep and goat Theileria and Babesia parasites. Parasitol Res 91, 398–406. doi.org/10.1007/s00436-003-0979-2
- Semenza JC, Lindgren E, Balkanyi L, Espinosa L, Almqvist MS, Penttinen P, Rocklöv J, 2016. Determinants and drivers of infectious disease threat events in europe. Emerg Infect Dis 22,

581-589. doi.org/10.3201/eid2204.151073

- Sillero N, 2011. What does ecological modelling model? A proposed classification of ecological niche models based on their underlying methods. Ecol Modell 222, 1343–1346. doi.org/10. 1016/j.ecolmodel.2011.01.018
- Stainforth DA, Chapman SC, Watkins NW, 2013. Mapping climate change in European temperature distributions. Environ Res Lett 8, 034031. doi.org/10.1088/1748-9326/8/3/034031
- Svensson J, Hunfeld KP, Persson KEM, 2019. High seroprevalence of Babesia antibodies among Borrelia burgdorferi-infected humans in Sweden. Ticks Tick-Borne Dis 10, 186–190. doi.org/10.1016/j.ttbdis.2018.10.007
- Tälleklint L, Jaenson TGT, 1998. Increasing geographical distribution and density of Ixodes ricinus (Acari: Ixodidae) in Central and Northern Sweden. J Med Entomol 35, 521–526. doi.org/10.1093/jmedent/35.4.521
- Tampieri MP, Galuppi R, Bonoli C, Cancrini G, Moretti A, Pietrobelli M, 2008. Wild ungulates as Babesia hosts in Northern and Central Italy. Vector-Borne Zoonotic Dis. 8, 667–674. doi.org/10.1089/vbz.2008.0001
- Vannier E, Krause PJ, 2012. Human babesiosis. N Engl J Med 366, 2397–2407. doi.org/10.1016/j.ijpara.2018.11.007
- Vor T, Kiffner C, Hagedorn P, Niedrig M, Rühe F, 2010. Tick burden on European roe deer (Capreolus capreolus). Exp Appl Acarol 51, 405–417. doi.org/10.1007/s10493-010-9337-0
- Walker A, 2003. Ticks of domestic animals in Africa. Edinburgh: Bioscience Reports.
- Wilson ML, Ducey AM, Litwin TS, Gavin TA, Spielman A, 1990. Microgeographic distribution of immature Ixodes dammini ticks correlated with that of deer. Med Vet Entomol 4, 151– 159. doi.org/10.1111/j.1365-2915.1990.tb00273.x
- Yabsley MJ, Shock BC, 2013. Natural history of Zoonotic Babesia: role of wildlife reservoirs. Int J Parasitol Parasites Wildl 2, 18–31. doi.org/10.1016/j.ijppaw.2012.11.003
- Zanet S, Bassano M, Trisciuoglio A, Taricco I, Ferroglio E, 2017. Horses infected by Piroplasms different from Babesia caballi and Theileria equi: species identification and risk factors analysis in Italy. Vet Parasitol 236, 38–41. doi.org/10.1016/j. vetpar.2017.01.003
- Zanet S, Trisciuoglio A, Bottero E, De Mera IGF, Gortazar C, Carpignano MG, Ferroglio E, 2014. Piroplasmosis in wildlife: Babesia and Theileria affecting free-ranging ungulates and carnivores in the Italian Alps. Parasite Vector 7, 1–7. doi.org/10.1186/1756-3305-7-70
- Zintl A, Finnerty EJ, Murphy TM, De Waal T, Gray JS, 2011. Babesias of red deer (Cervus elaphus) in Ireland. Vet Res 42, 1–6. doi.org/10.1186/1297-9716-42-7
- Zintl A, Mulcahy G, Skerrett HE, Taylor SM, Gray JS, 2003. Babesia divergens, a bovine blood parasite of veterinary and zoonotic importance. Clin Microbiol Rev 16, 622–636. doi.org/10.1128/CMR.16.4.622