

# Bacterial microbiota composition of *Ixodes ricinus* ticks: the role of environmental variation, tick characteristics and microbial interactions

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Ecological factors, host characteristics and / or interactions among microbes may all shape the occurrence of microbes and the structure of microbial communities within organisms. In the past, disentangling these factors and determining their relative importance in shaping within-host microbiota communities has been hampered by analytical limitations to account for (dis)similar environmental preferences ('environmental filtering'). Here we used a joint species distribution modelling (JSDM) approach to characterize the bacterial microbiome of one of the most important disease vectors in Europe, the sheep tick *Ixodes ricinus*, along ecological gradients in the Swiss Alps. Although our study captured extensive environmental variation along elevational clines, the explanatory power of such large-scale ecological factors was comparably weak, suggesting that tick-specific traits and behaviours, microhabitat and -climate experienced by ticks, and interactions among microbes play an important role in shaping tick microbial communities. Indeed, when accounting for shared environmental preferences, evidence for significant patterns of positive or negative co-occurrence among microbes was found, which is indicative of competition or facilitation processes. Facilitation occurred primarily among human pathogens, leading to co-infection within ticks, whereas competition was observed between the tick endosymbiont *Spiroplasma* and human pathogens. These findings highlight the important role of small-scale ecological variation and microbe-microbe interactions in shaping tick microbial communities and the dynamics of tick-borne disease.

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# 13 **Abstract**

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 15 occurrence of microbes and the structure of microbial communities within organisms. In the past,  
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 19 modelling (JSDM) approach to characterize the bacterial microbiome of one of the most  
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 23 suggesting that tick-specific traits and behaviours, microhabitat and -climate experienced by  
 24 ticks, and interactions among microbes play an important role in shaping tick microbial  
 25 communities. Indeed, when accounting for shared environmental preferences, evidence for  
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 31 microbial communities and the dynamics of tick-borne disease.

32 **Keywords:** tick-borne pathogens, species distribution modelling, community composition,  
 33 *Borrelia burgdorferi*, Lyme disease

## 34 Introduction

35 Microbial communities within organisms consist of symbionts, commensals, mutualists and  
 36 pathogens that co-occur simultaneously and potentially influence each other (Petney and  
 37 Andrews 1998; Rigaud et al. 2010; Sofonea et al. 2015). These microbial communities may be  
 38 shaped by a range of factors and processes, including the environment, host and microbe genetics  
 39 and the occurrence and abundance of other microbial species (Adair and Douglas 2017). For  
 40 example, certain microbial species might tolerate only specific biotic or abiotic conditions, which  
 41 makes it more likely that species with similar requirements co-occur within a host  
 42 ('environmental filtering', Dallas and Presley, (2014)). Similarly, the host's immune system can  
 43 influence colonization success of microbes (Hawley and Altizer 2011), with cross-immunity  
 44 preventing the colonization of different microbes with similar antigenic properties (Durand et al.  
 45 2015). Finally, direct interactions among microbes might affect colonization, or replication  
 46 success after colonization, through competition or facilitation processes. Competition may occur  
 47 when different microbes use the same, limited resources within a host (Lello et al. 2004),  
 48 whereas facilitation may occur directly through the production of public goods (West and  
 49 Buckling 2003) or indirectly through the modification of the host's physiology (Abraham et al.  
 50 2017) or immune defense (Rodríguez et al. 1999).

51 *Ixodes ricinus* is the most common tick species in Europe and an important vector for a range of  
 52 human and wildlife pathogens (Medlock et al. 2013). Its distribution and abundance are strongly  
 53 influenced by environmental conditions, in particular temperature and humidity (Cortinas et al.  
 54 2002; Gatewood et al. 2009). Previous studies on the bacterial community composition of *I.*  
 55 *ricinus* ticks have found approximately 100 OTUs (operational taxonomic units; Stackebrandt  
 56 and Goebel 1994) per tick (Greay et al. 2018). Most of them are environmental and free-living

bacteria but also several endosymbionts and human or wildlife pathogens were found, including *Borrelia* (Mannelli et al. 2012), *Rickettsia* (Venclikova et al. 2014), *Anaplasma* (Jahfari et al. 2014) and *Candidatus Neoehrlichia* (Kawahara et al. 2004).

Differences in the bacterial community structure and composition of ticks across habitats (Estrada-Peña et al. 2018), geographical sites (Carpi et al. 2011), and tick life stages and sexes (Carpi et al. 2011; Vayssier-Taussat et al. 2013) have been documented. Large-scale biotic or abiotic factors such as vegetation structure, elevation, temperature or rainfall may influence tick microbial communities directly, or indirectly through effects on tick physiology or activity patterns (van Treuren et al. 2015) or via influencing the distribution and abundance of tick hosts species (Randolph et al. 1999; MacDonald et al. 2017). Small-scale and/or tick-specific effects on microbial communities may be explained by microhabitat or microclimatic conditions experienced by individual ticks (Gern et al. 2008), individual tick behavior or genetics (Hawlena et al. 2013), direct interactions among microbes (Moutailler et al. 2016) or parallel acquisition from a host during a bloodmeal (Andersson et al. 2014; Belli et al. 2017; Swei and Kwan 2017).

Currently, the relative importance of these factors in shaping tick microbial communities is not well understood, which hampers progress in our understanding of the processes shaping microbial communities in nature and predicting the occurrence of specific microbes (e.g., human pathogens). Elevational gradients are excellently suited to quantify the importance of large-scale ecological variation in shaping tick bacterial microbiota because they cover a large range of environmental conditions within a small geographical area. Furthermore, including replicated transects along gradients allow us to quantify the robustness of ecological associations within sites and along elevational clines on tick microbial communities.

*Ixodes* ticks are commonly found to be co-infected with several human or wildlife pathogens (Andersson et al. 2013; Michelet et al. 2014; Diuk-Wasser et al. 2016; Moutailler et al. 2016). Currently, it is unknown whether these co-infection patterns are caused by similar environmental preferences of pathogens, parallel acquisition from host communities or direct microbe-microbe interactions within ticks. Yet, previous studies suggest that the latter process, (i.e., facilitation and competition processes among microbes) may play a role in shaping the microbial communities (Haine 2008; Bonnet et al. 2017). For example, it has been found that pathogenic *Rickettsia* species prevent co-infection with other *Rickettsia* species in *Dermacentor variabilis* ticks (Macaluso et al. 2002), whereas the presence of *Francisella* sp. endosymbionts increases the colonization success of pathogenic *Francisella novicida* in *D. andersoni* ticks (Gall et al. 2016). Facilitation has also been suggested to promote co-infection with different *Borrelia afzelii* strains in *Ixodes ricinus* ticks (Andersson et al. 2013). Most strikingly, dysbiosis in *I. scapularis* ticks (i.e., ticks with low microbiotal diversity) leads to a defective peritrophic matrix which decreases the colonization success of *B. burgdorferi* s.s., suggesting that the pathogen requires the presence of an intact microbiota to be able to invade ticks (Narasimhan et al. 2014). Thus, the microbial community may have a crucial impact on vector competence of ticks and thereby on disease dynamics.

Yet, as outlined above, co-occurrence of microbes can be due to environmental filtering or direct microbial interactions, and distinguishing between these processes is non-trivial. Indeed, previous studies that have documented pathogen co-occurrence in ticks have not accounted for potential confounding variables such as shared ecological requirements, and are thus limited in their ability to differentiate between accidental co-occurrences due to shared environmental niches, and co-occurrence shaped by facilitation or competition among microbes.

To address these gaps, we exploited the substantial environmental heterogeneity along replicated elevational gradients in the Swiss Alps to quantify the relative importance of environmental factors, tick characteristics and direct microbial interactions in influencing the structure of bacterial communities in *I. ricinus* ticks in general, and the (co-)occurrence of human or wildlife pathogens in particular, using a combination of 16S sequencing and joint species distribution modelling (JSDM) (Ovaskainen et al. 2015; Warton et al. 2015). By taking shared environmental preferences into account, JSDMs allows to identify residual co-occurrence patterns among microbes that can result from unaccounted environmental effects or direct microbial interactions. Specifically, we ask (i) how do large-scale abiotic factors and small scale tick-level variables affect tick microbiota composition, (ii) which large-scale abiotic and small-scale tick-level variables predict pathogen occurrence, and (iii) are there patterns of non-random microbial co-occurrence that cannot be explained by environmental responses, implying unmeasured variables, such as microbial interactions.

## Materials and methods

### *Tick sampling*

Questing *Ixodes ricinus* ticks were collected at three locations in the Swiss Alps (Kanton Graubünden). At each location, one site at low (630 - 732 m above sea level, masl), one at medium (1 094 – 1 138 masl) and one at high (1 454 – 1 673 masl) elevation were identified (Fig. 1, Table 1, N = 9 sampling sites). At each site, questing ticks were sampled thrice, once in June, once in July, and once in August 2014 by dragging a white blanket (1 m x 1 m) over the ground vegetation as described previously (Lemoine et al. 2018). Ticks were collected from the

blanket and stored in 95% ethanol. Tick species, sex and life stage were verified by morphological features following (Hillyard 1996) using a stereomicroscope.

# *Environmental variables*

For each sampling site, we compiled information on large-scale, site-level ecological variables by obtaining data on elevation, slope and aspect using DHM25, land use data from swissTLM3D (both from Federal Office of Topography swisstopo) and data on temperature and precipitation from Landscape Dynamics (Swiss Federal Research Institute for Water, Snow and Landscape Research WSL and Federal Office of Meteorology and Climatology MeteoSwiss, (Thornton et al. 1997)). Data on *I. ricinus* abundance and the abundance of a key tick host, the bank vole (*Myodes glareolus*), as well as the ratio of bank vole to other rodents at our sampling sites were obtained from (Cornetti et al. 2018). Details on the different variables and a justification why they were included to characterise large-scale ecological conditions is provided in the Supplementary Material.

# *Tick DNA isolation and quantification of neutral genetic diversity*

The number of analysed ticks per site is presented in Table 1. Although we aimed to include similar numbers of *I. ricinus* ticks from each sampling site and month, variation in the number of ticks per site (Table 1) was unavoidable because of variation in tick abundance across sites (Lemoine et al. 2018). To avoid environmental contamination, we performed DNA isolation and amplifications in a laminar flow cabinet. Each tick was washed thrice with sterile water before sterilizing it with 3% hydrogen peroxide. Ticks were then cut in half with a sterilized blade to



facilitate DNA isolation. DNA was extracted using DNeasy Blood & Tissue kit (Qiagen; Hilden, Germany).

Host genetics may affect pathogen and endosymbiont colonisation and replication success (Archie and Ezenwa 2011). In order to quantify individual and population-level genetic diversity, we genotyped ticks at 11 microsatellite markers in two multiplexed amplifications (see Supplementary Material for details). Not all markers were successfully amplified in all samples, but none of the samples contained more than two failed markers. We used package *poppr* (Kamvar et al. 2014) in R 3.4.1 (Team 2013) to test for linkage disequilibrium and deviation from Hardy-Weinberg equilibrium. Individual observed heterozygosity was determined for each tick as a proportion of heterozygous markers to all successfully amplified markers. Expected population level heterozygosity was determined with *poppr*. The former was used as a tick-level explanatory variable (together with tick sex and life stage), the latter was used as a site-level explanatory variable.

### *Tick microbiota sequencing*

We quantified tick bacterial community composition by sequencing the hypervariable V3-V4 region of the 16S rRNA (16S) gene. Negative controls (extraction reagent blank, N=2 and PCR controls, N=3) were processed alongside the tick samples. Sequencing libraries were prepared following the Earth Microbiome 16S Illumina Amplicon protocol, using the primers 515FB and 806RB (Carey et al. 2013) (see Supplementary Material for details). Samples and negative controls were randomized across two plates. The libraries were sequenced on Illumina MiSeq at the Functional Genomic Center Zurich with a target length of 250 bp following the

manufacturer's protocol. The obtained sequence data were analyzed following the *mothur* pipeline with MiSeq standard operation procedures (Kozich et al. 2013). Sequences have been deposited to the Sequence Read Archive under BioProject PRJNA506875. The complete metadata of the samples and their matching sequence accession numbers have been submitted to FigShare (doi: 10.6084/m9.figshare.7380767).

A special focus of our analysis was on tick endosymbionts and tick-borne human or wildlife pathogens (Table 2), which are obligate residents. Identification of endosymbionts and pathogens is described in Supplementary Material.

#### *Joint species distribution modelling of microbiota composition*

Only samples with > 500 amplicons and OTUs which were present in at least two samples were included in the analyses (Table 1). As the most common OTU, the intra-mitochondrial endosymbiont *Candidatus* Midichloria (Lo et al. 2006), was present in all samples, it was not included in the modelling. For the occurrence matrix, an OTU was determined to be present in a tick if >5 amplicons were identified in a sample (Aivelo and Norberg 2017).

We used a JSDM framework called Hierarchical Modelling of Species Communities (HMSC, Ovaskainen et al. 2017a) to examine how environmental variables correlate with human or wildlife pathogen and tick endosymbiont occurrence in ticks, and whether there are non-random residual associations among different OTUs and/or oligotypes, implying direct facilitation or competition effects among microbes. This approach combines information on environmental covariates, bacterial species traits, spatiotemporal context and sampling design to explain the presence or absence of OTUs (Fig. S1). The associations among OTUs are captured with the

latent part of the framework, modelling the residual variance after accounting for the effects of the environment with latent variables. The estimates for these latent variables can be then translated into residual correlations among response variables, i.e. OTUs and/or oligotypes. These correlations thus reflect (dis)associations which cannot be explained by shared responses to the environment.

We compiled occurrence matrices for OTUs for each individual tick as a response variable. For each sampling unit, i.e. a row in our response variable matrix, we included information on the identity of the sampling unit (tick ID), its location, sampling site and month, describing the study design. To reach a better resolution within specific OTUs, we analyzed known human or wildlife pathogens, tick endosymbionts and their close relatives within the 100 most common OTUs with oligotyping pipeline (Eren et al. 2014). Oligotyping uses all the sequences, which form an OTU, and performs Shannon Entropy Analysis to regroup sequences based on within-OTU variation. This results in higher-resolution grouping than OTUs as the different oligotypes might differ only by a single nucleotide (Eren et al. 2014). We used the standard operation procedures of the oligotyping pipeline software (Eren et al. 2013).

Including a large number of explanatory variables in statistical models is inherently challenging. To reduce the number of variables, while maintaining their information value, we used two variable sets in the model: a) a set of full-effect explanatory variables, and b) explanatory variables under variable selection (Ovaskainen et al. 2017b). The full-effect variable set included an intercept, two tick-level variables (tick sex or life stage and individual heterozygosity) and two site-level variables (tick abundance and elevation of the sampling site). Additionally, we included information whether a specific OTU is an endosymbiont and/or a human or wildlife pathogen (Abrego et al. 2016). This allowed us to test if endosymbionts and/or human or wildlife

pathogens respond differentially to environmental conditions than other OTUs. The set of explanatory variables under variable selection included additional information on the environmental conditions of the sites (namely the number of days above 7 °C during the year, monthly precipitation, mean monthly temperature, forest coverage, slope, aspect, bank vole abundance, the proportion of voles to other rodents and expected tick heterozygosity) (Table S1). We considered all parameter estimates, including associations among bacterial OTUs, having strong statistical support and thus being “significant” if the 90% central credible interval of the parameter did not overlap with zero (following Aivelo and Norberg 2017; see Supplementary Material for additional model details).

Although JSDM is a powerful approach to model community structure, it has a number of limitations. First, it assumes that interactions among microbes are similar across environments (but see Tikhonov et al. 2017). This is not necessarily the case as both abiotic and biotic factors may shape microbial interactions (Elliot et al. 2002). Second, the model assumes that the explanatory variables affect the microbial community composition (or rather, the presence or absence of individual OTUs), but not vice versa. However, this is a valid assumption for most environmental (e.g. elevation and temperature) and tick-related variables (e.g. tick sex, life stage) included in our models. Thirdly, covariation among explanatory variables poses a problem to any correlative modelling approach. Our model is built on two distinct variable sets to aid in handling such covariation: the full variable set includes elevation, whereas the variables with the strongest covariation (i.e., temperature and precipitation) are included in the variable selection set.

## Results

# 234 *Ixodes ricinus* microbiota composition

235 We 16S sequenced the bacterial community of 92 *Ixodes ricinus* ticks which resulted in 13 214  
 236 477 amplicons. No amplification was observed in the negative controls (i.e., their sequencing did  
 237 not result in any sequences). After contig assembly and quality control 1 656 287 sequences were  
 238 retained. Most of the discarded amplicons were either shorter than 250 bp or chimeras. There  
 239 was a median of 1 562 quality-controlled amplicons per sample, with an interquartile range of 6  
 240 319. 82 samples with more than 500 amplicons per sample, a plateauing accumulation curve and  
 241 a Good's coverage estimator  $\geq 0.95$  were included in the subsequent analyses. In total, 5 181  
 242 bacterial OTUs were identified. The median number of OTUs when rarified to 500 amplicons  
 243 per sample was 89 OTUs, with a 95% confidence interval of 78.3 - 98.5 OTUs.

244 Six OTUs were present in at least 90% of the samples: *Ca. Midichloria* (Otu0001),  
 245 *Sphingomonas* (Otu0002, 0006 and 0007), *Pseudomonas* (Otu0011) and *Delftia* (Otu0012).  
 246 Together, they represented 50.2 % of all amplicons. We used oligotyping to further divide  
 247 OTU0031 '*Rickettsia*' into two oligotypes labelled as '*R. helvetica*' and '*R. monacensis*', and  
 248 OTU0086 '*Borrelia*' into four oligotypes labelled as '*B. afzelii*', '*B. valaisiana*' and '*B. garinii*'  
 249 and '*B. miyamotoi*'. After excluding rare OTUs, 635 OTUs and oligotypes were used in  
 250 subsequent analyses, including 14 endosymbionts and / or human or wildlife pathogens (Table  
 251 2).

252

# 253 *Tick microbiota variance partitioning*

254 Variance partitioning revealed that most of the variation in tick microbiota composition  
 255 explained by our model related to tick ID: for the hundred most common OTUs, tick ID

accounted for 64.1% of the variation explained by the model. Fixed effects (e.g., tick life stage, elevation; see Table S1) accounted for 12.5% (tick-level: 7.3%, site-level: 5.2%) and spatial and temporal random effects (i.e., location, site and month) explained 23.3% (Fig. 2). This suggests that there is extensive tick-level variation which cannot be accounted for by tick-specific characteristics included in our model (i.e., sex, life stage, genetic diversity) or site-level environmental factors. The pattern differed slightly for endosymbionts and human or wildlife pathogens: while tick ID was still the most important variable explaining 49.9%, fixed effects explained 31.8% (tick-level: 17.5%, site-level: 14.3%) and spatial and temporal random effects explained 18.3% of the total variation explained by the model, when averaged over all pathogens and endosymbionts (Fig. 2).

#### *Tick-specific and environmental factors related to OTU occurrence*

The occurrence of tick endosymbionts and human or wildlife pathogens was strongly associated with specific explanatory variables, yet associations were typically microbe-specific rather than universal (Table 3). The two most important variables explaining the presence or absence of tick endosymbionts and human or wildlife pathogens were tick sex and elevation of the sampling site: adult female ticks were less likely to harbour the endosymbionts *Spiroplasma*, *Rickettsiella*, *Lariskella* and *Rickettsia* spp. (Table 3), and ticks at higher elevations had higher probability to harbour *R. helvetica* and *R. monacensis*, but were less probable to harbour *B. garinii* (Table 3). Topography was also a significant predictor of tick endosymbionts and human or wildlife pathogen occurrence, with ticks from sites facing northwards having a higher probability of harbouring *Spiroplasma* and *B. afzelii*, and ticks from sites on steeper slopes having a higher probability of harbouring *Rickettsia* sp. (Table 3). Furthermore, a higher tick density was

associated with a higher probability of *Rickettsiella* and *Ca. Neoehrlichia* occurrence (Table 3). Relationships between tick life stage, mean temperature, the number of days > 7 °C or forest cover and the occurrence of specific OTUs were not strongly statistically supported.

The effect sizes of strongly statistically supported associations varied substantially (Fig. S3a-i). For example, threefold increase in vole abundance corresponded to less than one percentage point decrease of *R. monacensis* prevalence (Fig. S3b), whereas a threefold increase in tick abundance corresponded to a threefold increase in *Neoehrlichia* prevalence from 8% to 27% (Fig. S3e).

#### *Patterns of microbial association within ticks*

Numerous bacterial OTUs were either significantly more or less likely to co-occur within a tick than expected by chance after accounting for shared environmental preferences (Fig. 3a; Table S5). At the level of the individual tick, the occurrence of the tick endosymbiont *Spiroplasma* was negatively associated with the occurrence of the endosymbiont *Lariskella* and several tick-borne pathogens, namely *Rickettsia* sp., *Ca. Neoehrlichia* and *B. miyamotoi* (Fig. 3a). Associations among pathogens, if they occurred, were all positive (Fig. 3a), suggesting that ticks are more likely to be co-infected with several human or wildlife pathogens simultaneously than expected by chance or based on shared environmental preferences. *Borrelia* oligotypes showed positive co-occurrence patterns among each other, except for *B. miyamotoi*, which was not associated with other *Borrelia* sp., but negatively with *Spiroplasma* and positively with *Lariskella*. At the level of the sampling site, significant associations were sparser. *Spiroplasma* was more likely to co-occur with *Lariskella* and *Rickettsiella* across sites, whereas the latter two were less likely to co-occur across sites than expected by chance after accounting for shared environmental preferences (Fig. 3b).

302

# 303 Discussion

304 We used a JSDM framework to quantify the relative importance of large scale, site-level  
 305 environmental variables, tick-level characteristics and interactions among microbes in shaping  
 306 tick microbiota composition along elevational gradients in the Swiss Alps. We show that  
 307 although our study captured extensive environmental variation, with sampling sites spanning  
 308 across an elevational gradient from 630 – 1 580 masl, and a large number of ecological variables  
 309 was considered in our models, the explanatory power of such large-scale ecological factors was  
 310 comparably weak. In contrast, individual tick ID explained over 60% of the variation in  
 311 microbiota composition. This substantial microbiota variation across individual ticks may be  
 312 partly explained by some of the bacteria present in ticks being non-resident (i.e., bacteria that  
 313 were accidentally obtained from the environment). Indeed, there has been a debate whether ticks  
 314 have a stable microbiota (Ross et al. 2017), mirroring the wider debate on how common resident  
 315 microbiota is in arthropod hosts (Hammer et al. 2017).

316 However, also for endosymbionts and human or wildlife pathogens, which are obligate resident,  
 317 tick ID accounted for half of the variation in occurrence, suggesting that microhabitat or -  
 318 climatic conditions experienced by individual ticks, tick-specific traits and behaviors not  
 319 included in our models, as well as microbial interactions within ticks such as facilitation and  
 320 competition (Abraham et al. 2017; Gurfield et al. 2017), play a crucial role in shaping microbiota  
 321 composition and the occurrence of endosymbionts and human or wildlife pathogens in *I. ricinus*.  
 322 Focusing on such small-scale variables, rather than large-scale climatic or environmental factors  
 323 as is usually done when modelling tick-borne pathogen prevalence (Norman et al. 2016; Rosà et



al. 2018), is thus likely a more fruitful approach to advance our understanding of microbiota composition of natural populations as well as (tick-borne) disease dynamics.

Co-occurrence of human or wildlife pathogens in ticks has been documented previously, both in *I. ricinus* (Lommano et al. 2012; Michelet et al. 2014) and other tick species (Gurfield et al. 2017; Laaksonen et al. 2018). Yet, previous studies did not control for shared environmental preferences (i.e., environmental filtering), which limited their ability to disentangle shared responses to the environment from direct microbe-microbe interactions. Our study revealed that when accounting for shared environmental preferences, associations among human or wildlife pathogens were often pronounced and mostly positive. These positive associations may result from direct facilitation among microbes or parallel colonization from co-infected tick hosts.

Within ticks, the strong positive associations among the Lyme disease-causing *Borrelia* genospecies (*B. afzelii*, *B. garinii* and *B. valaisiana*) were particularly striking. This positive co-occurrence is surprising because *B. garinii* and *B. valaisiana* are bird specialists (Hanincova et al. 2003b; Comstedt et al. 2011), whereas *B. afzelii* is a rodent specialist (Hanincova et al. 2003a). Thus, the parallel colonization from co-infected tick hosts cannot explain this pattern. Rather the positive co-occurrence is indicative of facilitation processes among *Borrelia* genospecies, as has been suggested previously (Andersson et al. 2013). Such facilitation, and the resulting co-infection of ticks with several *Borrelia* genospecies has implications for the severity, diagnosis, treatment and control of Lyme disease. Finally, the co-occurrence of these different *Borrelia* genospecies suggests that *I. ricinus* feeds on multiple, phylogenetically diverse host species during its life cycle and does not show pathogen-mediated host specialization as has been suggested previously (McCoy et al. 2005, 2013).

Although associations among microbes were mostly positive, there were negative associations between the tick endosymbiont *Spiroplasma* and several human or wildlife pathogens, which may be explained by competition. The most common infection route for *Spiroplasma* is maternal (i.e., vertical) transmission (Herren and Lemaitre 2011), indicating that horizontal or environmental transfer plays a minor role in its transmission. Protective effects of *Spiroplasma* have been previously described in *Drosophila* spp., where *Spiroplasma* is associated with a decreased probability of nematode and parasitoids infections (Xie et al. 2010; Jaenike et al. 2013). Although the exact mechanisms mediating *Spiroplasma*-induced competition effects are currently unknown, this finding may stimulate further research into the potential of tick endosymbionts to manage tick-borne pathogens.

In contrast to the numerous positive or negative associations among microbes at the tick-level, little statistical support for positive or negative microbial co-occurrence was found at the site-level, with the exception of the associations among three endosymbionts. Interestingly, the pattern of co-occurrence of *Spiroplasma* and *Lariskella* at the site-level was opposite from what was observed at the tick-level. It suggests that *Spiroplasma* and *Lariskella* are more likely to co-occur at the same sites but not within the same tick. Co-occurrence at the site-level is most likely due to an environmental variable not included in our model, for which the three OTUs had similar responses. Importantly, the substantial reduction of microbe-microbe associations at the site level highlights relevant environmental variation across sites was captured in our models, strengthening our conclusion that the patterns of positive or negative co-occurrence among microbes observed at the tick-level can be due to direct microbial interactions.

Despite the large among-tick variation in microbiota composition, we identified a range of environmental variables that significantly predicted the occurrence of specific tick

endosymbionts and human or wildlife pathogens. However, the predictor variables as well as their effect were typically OTU-specific rather than universal. For example, *B. garinii* was less likely to occur at higher elevations, whereas *R. helvetica* and *R. monascensis* were more likely to occur at higher elevations. Generally, the environmental factors shaping *Rickettsia* spp. distribution are poorly understood, as is their range of host species (Halos et al. 2010; Ereemeeva and Dasch 2015). Yet, it has previously been found that spotted fever incidence in humans, caused by *R. rickettsii*, is highest in areas or regions, where ticks are less common (Atkinson et al. 2013). This is in line with our findings and suggests that *Rickettsia* spp. are more likely to colonize ticks living under suboptimal conditions (e.g. at range edges).

The finding that *B. garinii* is less likely to occur at higher elevations is in line with previous observations (Jouda et al. 2004b; Cornetti et al. 2018) and may be explained by changes in vegetation structure and associated changes in host communities (Halos et al. 2010), in particular changes in the diversity and/or abundance of birds, the natural hosts of *B. garinii* (Comstedt et al. 2011). In contrast, the occurrence of the mammal specialist *B. afzelii* was not related to elevation, potentially because elevational clines in mammal diversity and/or abundance are less pronounced (McCain 2005). Indeed, we did not observe an association between elevation and bank vole abundance across our study sites (ANOVA:  $F_{1,8}=0.357$ ,  $p = 0.57$ ,  $R^2 = 0.05$ ).

Interestingly, temperature and precipitation, which vary strongly across elevational gradients (average temperature and precipitation: high elevation sites: 11.8 °C and 17.8 mm per month; in low sites: 16.5 °C and 12.1 mm per month), were not significant predictors of the occurrence of endosymbionts or human or wildlife pathogens, with the exception of precipitation correlating positively with the probability of *Rickettsiella* occurrence. This may be partly explained by the temperature and precipitation measures included in our models not fully capturing the

microclimatic variation across sites and along elevational clines. Indeed, slope and aspect, which are important determinants of the topography, and thus microclimate (Bennie et al. 2008), were significant predictors of pathogen and endosymbiont occurrence. The probability of *Rickettsia* sp. occurrence was higher on steeper slopes. Furthermore, the probability of occurrence was higher on north-facing slopes for *B. afzelii* and *Spiroplasma* and higher on south-facing slopes for *Rickettsiella* (see also (Stuen et al. 2013)). Microclimatic conditions may affect microbial occurrence directly, or indirectly via affecting tick behavior or host community composition (Swei et al. 2011; Lawson et al. 2014). Furthermore, topography can affect population connectivity and dispersal in metapopulation networks (Swei and Kwan 2017).

Previous work has found that tick abundance is a strong predictor of *Borrelia* spp. prevalence, potentially because larger tick populations facilitate co-feeding transmission (Jouda et al. 2004a). No relationship between *Borrelia* spp occurrence and tick abundance was observed in our study. However, both *Ca. Neoehrlichia* and *Rickettsiella* were more common at sites where ticks were more abundant, suggesting that co-feeding transmission may also play a role in the life cycle of these microbes.

Finally, differences in host competence can lead to dilution effects and thus affect the prevalence of tick-borne pathogens (Keesing et al. 2006). Whereas for some tick-borne pathogens the vertebrate hosts are known or suspected (e.g. small mammals for *B. afzelii* (Hanincova et al. 2003a) and *Ca. Neoehrlichia* (Jahfari et al. 2012), birds for *B. garinii* and *B. valaisiana* (Hanincova et al. 2003b), both for *Anaplasma* (Keesing et al. 2012) and *R. helvetica* (Sprong et al. 2009)), for others the host species range is less well understood (e.g. *B. miyamotoi* (Wagemakers et al. 2015)). The bank vole is a common tick host at our study sites and their abundance was a significant negative predictor of *R. monacensis* and *R. helvetica* occurrence.

Interestingly, bank voles are not known hosts for either (Burri et al. 2014). Most likely, the relation is thus indirect, explained by an unmeasured biotic or abiotic variable that correlates with bank vole abundance. No evidence was found that the proportion of bank voles to other rodents affects the prevalence of tick-borne pathogens.

In conclusion, our study demonstrates that a JSDM framework can contribute to a better understanding of the factors shaping bacterial communities in natural populations as well as patterns of co-occurrence among microbes. Overall, our study highlights the role of small-scale, tick-level characteristics rather than large-scale ecological variation in shaping microbial communities of *I. ricinus*. We identified a number of ecological variables that predict the occurrence of specific tick endosymbionts and human or wildlife pathogens with strong statistical support, but these effects were typically microbe-specific rather than universal. This highlights that environmental change can have different, even opposite effects on different human pathogens, and thus disease risk. Furthermore, by accounting for shared environmental preferences, our approach identified patterns of microbial co-occurrence that are consistent with microbe-microbe interactions, which result in pathogen co-infections within ticks, as well as competition between *Spiroplasma* and a number of human or wildlife pathogens. The latter opens up new and exciting avenues for the control and management of tick-borne diseases in regions with high human disease incidence.

### Acknowledgements

We thank Mélissa Lemoine for providing ticks, the numerous people who helped collecting ticks in the field, Glauco Camenisch, Elisa Granato, Jennifer Morger and Alessia Pennachia for help with laboratory work, Lucy Poveda for help with MiSeq sequencing, Frédéric Guillaume for providing access to IT infrastructure, Janine Bolliger and Dirk Schmatz from Swiss Federal

Institute for Forest, Snow and Landscape Research WSL for providing spatial data and for help with spatial analyses, and Otso Ovaskainen for discussions on the modelling.

# **Funding statement**

This study was funded by Finnish Cultural Foundation Postdoc Pool grant (to TA), the Stiftung für wissenschaftliche Forschung an der Universität Zürich (17\_027), the Swiss National Science Foundation (PP00P3\_128386 and PP00P3\_157455), the University of Zurich Research Priority Program “Evolution in Action: from Genomes to Ecosystems”, the Faculty of Science of the University of Zurich, and the Baugarten Stiftung (all to BT).

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# Figure legends

Figure 1: Location of tick sampling sites in the Swiss Alps. Different colours of circles (i.e., black, grey, white) represent the different locations. Rivers and motorway are shown in black. Modified from (Cornetti et al. 2016).

Figure 2: Tick microbial community variance partitioning for different fixed and random effects. The first three columns represent tick endosymbionts, the next three columns are OTUs which are both tick endosymbionts and human pathogens and the subsequent six columns represent human pathogens. The other columns represent the 88 most common OTUs found in *I. ricinus*, ordered by amplicon frequency. Month, sampling site, location and tick ID were included in the model as random effects, whereas fixed effects were divided into environmental (elevation, temperature, precipitation, forest coverage, slope, aspect, vole abundance and vole-to-other-rodents ratio) and tick-specific variables (life stage or sex, individual heterozygosity, abundance, expected population heterozygosity). See raw data in Figshare for information on OTU labels.

Figure 3: Residual association patterns among endosymbionts and human pathogens within ticks on a) individual tick-level and b) on site-level after accounting for shared environmental preference. Red lines represent positive associations and blue lines negative associations. Only associations with strong statistical support (i.e., based on the 90% central credible interval) are presented. Darker colors indicate stronger associations.

**Table 1** (on next page)

Tick sampling sites in the Swiss Alps.



1 Table 1: Tick sampling sites in the Swiss Alps.

Locatio n	Site	Coordinates		Elevation (masl)	Succesfully sequenced <i>Ixodes ricinus</i> ticks		
		North	East		nymphs	males	females
1	Sagogn	46.783	9.233	693	0	9	15
	Flims	46.827	9.280	1138	3	5	3
	Ruschein	46.795	9.169	1454	0	1	1
2	Rodels	46.760	9.425	630	2	5	4
	Tomils	46.772	9.453	1144	3	6	4
	Feldis	46.789	9.453	1673	1	1	0
3	Passug	46.840	9.538	732	0	5	6
	Castiel	46.826	9.569	1094	0	3	3
	Praden	46.817	9.589	1582	1	0	1

2

**Table 2**(on next page)

Common tick endosymbionts and/or putative human pathogens observed in *I. ricinus* ticks.

See Supplementary Materials for information on OTU assignment.

- 1 Table 2: Common tick endosymbionts and/or putative human pathogens observed in *I. ricinus* ticks. See
- 2 Supplementary Materials for information on OTU assignment.

OTU	Label	Human pathogen / tick endosymbiont	Occurrence (% of analyzed ticks)
Otu0001	<i>Midichloria</i>	endosymbiont	100
Otu0003	<i>Spiroplasma</i>	endosymbiont	41
Otu0005	<i>Rickettsiella</i>	endosymbiont	63
Otu0021	<i>Lariskella</i>	endosymbiont	49
Otu0031	<i>Rickettsia helvetica</i>	both	16
	<i>R. monacensis</i>	both	6
Otu0067	<i>Rickettsia sp.</i>	both	25
Otu0076	<i>Anaplasma</i>	both	33
Otu0086	<i>Candidatus</i> <i>Neoehrlichia</i>	both	22
Otu0088	<i>Borrelia afzelii</i>	pathogen	9
	<i>B. miyamotoi</i>	pathogen	10
	<i>B. garinii</i>	pathogen	6
	<i>B. valaisiana</i>	pathogen	2

3

4

# Table 3 (on next page)

Associations between tick-specific and environmental variables and the occurrence of endosymbionts and human pathogens in *I. ricinus* ticks.

A positive sign indicates that higher variable values are associated with a higher probability of OTU occurrence. A higher aspect value means that a site is facing northwards. Only associations with strong statistical support (based on the 90% central credible interval) are presented.

1 Table 3. Associations between tick-specific and environmental variables and the occurrence of  
 2 endosymbionts and human pathogens in *I. ricinus* ticks. A positive sign indicates that higher variable  
 3 values are associated with a higher probability of OTU occurrence. A higher aspect value means that a  
 4 site is facing northwards. Only associations with strong statistical support (based on the 90% central  
 5 credible interval) are presented.

Full variable set		Variable selection set													
		Tick sex (Female)	Tick life stage (Nymph)	Tick abundance	Tick heterozygosity	Elevation	Tick population expected heterozygosity	Number of days > 7C°	Precipitation	Mean temperature	Forest cover	Slope	Aspect	Vole abundance	Vole/ other rodents ratio
Otu0003	<i>Spiroplasma</i>	—											+		—
Otu0005	<i>Rickettsiella</i>	—		+	—		—		+				—		
Otu0022	<i>Lariskella</i>	—			—										
Otu0031	<i>Rickettsia helvetica</i>					+								—	
	<i>R. monacensis</i>					+	—							—	
Otu0067	<i>Rickettsia sp.</i>	—										+			
Otu0076	<i>Anaplasma</i>														
Otu0086	<i>Ca. Neoehrlichia</i>			+											
Otu0088	<i>Borrelia afzelii</i>												+		
	<i>B. miyamotoi</i>														
	<i>B. garinii</i>					—									

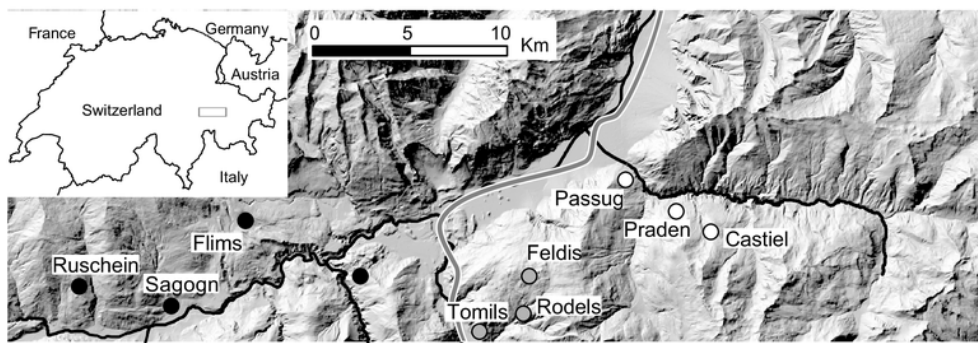
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*B. valaisiana*

# Figure 1

Location of tick sampling sites in the Swiss Alps.

Different colours of circles (i.e., black, grey, white) represent the different locations. Rivers and motorway are shown in black. Modified from (Cornetti et al. 2016).

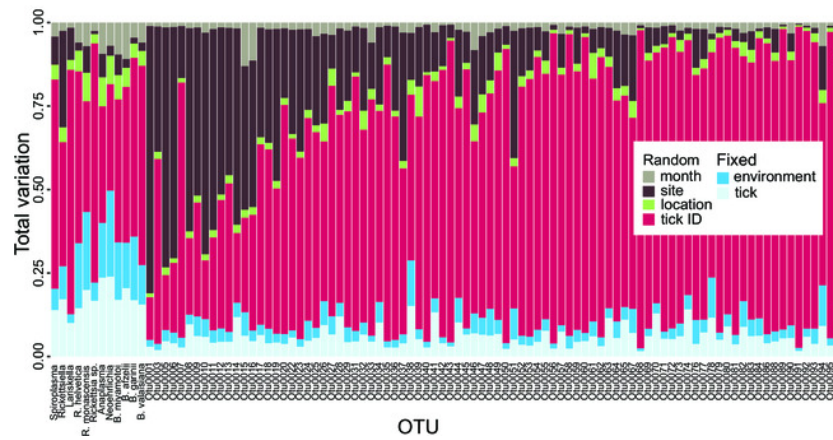


# Figure 2

Tick microbial community variance partitioning for different fixed and random effects.

The first three columns represent tick endosymbionts, the next three columns are OTUs which are both tick endosymbionts and human pathogens and the subsequent six columns represent human pathogens. The other columns represent the 88 most common OTUs found in *I. ricinus*, ordered by amplicon frequency. Month, sampling site, location and tick ID were included in the model as random effects, whereas fixed effects were divided into environmental (elevation, temperature, precipitation, forest coverage, slope, aspect, vole abundance and vole-to-other-rodents ratio) and tick-specific variables (life stage or sex, individual heterozygosity, abundance, expected population heterozygosity). See raw data in Figshare for information on OTU labels.





# Figure 3

Residual association patterns among endosymbionts and human pathogens within ticks on a) individual tick-level and b) on site-level after accounting for shared environmental preference.

Red lines represent positive associations and blue lines negative associations. Only associations with strong statistical support (i.e., based on the 90% central credible interval) are presented. Darker colors indicate stronger associations.

