1	Global spread of helminth parasites at the human – domestic animal –
2	wildlife interface
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4	Running head: Parasites at the human–animal interface
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29

30 Abstract

Changes in species distributions open novel parasite transmission routes at the human– wildlife interface, yet the strength of biotic and biogeographical factors that prevent or facilitate parasite host shifting are not well understood.

We investigated global patterns of helminth parasite (Nematoda, Cestoda, Trematoda) 34 sharing between mammalian wildlife species and domestic mammal hosts (including 35 36 humans) using > 24,000 unique country-level records of host-parasite associations. We used 37 hierarchical modelling and species trait data to determine possible drivers of the level of parasite sharing between wildlife species and either humans or domestic animal hosts. We 38 39 found the diet of wildlife species to be a strong predictor of levels of helminth parasite 40 sharing with humans and domestic animals, followed by a moderate effect of zoogeographical region and minor effects of species' habitat and climatic niches. Combining 41 model predictions with the distribution and ecological profile data of wildlife species, we 42 projected global risk maps that uncovered strikingly similar patterns of wildlife parasite 43 sharing across geographical areas for the different domestic host species (including humans). 44 These similarities are largely explained by the fact that widespread parasites are commonly 45 recorded infecting several domestic species. 46

47 If the dietary profile and position in the trophic chain of a wildlife species largely drives its
48 level of helminth parasite sharing with humans/domestic animals, future range shifts of host
49 species that result in novel trophic interactions may likely increase parasite host shifting and
50 have important ramifications for human and animal health.

52 **Introduction**

53 The emergence of parasitic diseases is largely a consequence of the exploitation of novel host species by parasites capable of shifting hosts (Lloyd-Smith et al., 2009). A central goal in 54 55 disease ecology is thus to identify factors that enable parasite sharing, especially since determinants of parasite sharing can influence the spread of parasites to new habitats and 56 biogeographic regions. For zoonotic diseases (i.e. infectious diseases of humans caused by 57 58 parasites that have an animal reservoir) a key determinant of emergence is overlapping 59 environmental conditions and biological traits that enable parasites to be shared by human and animal hosts. Along early human migration pathways, increased physical contact with 60 61 endemic animal and plant species led to increased exposure to novel parasites (Pedersen & 62 Davies, 2009; Pulliam, 2008), especially those acquired through ingestion of wild animal meat (Reinhard et al., 2013). Anthropogenic land use, conversion of natural habitats into 63 64 production landscapes, and intensification of international travel and wildlife trades continue to diminish or shift former geographical barriers between humans and wildlife, likely 65 facilitating exposure to novel parasite pools (Murray et al., 2015; Patz et al., 2008). In 66 contrast, decreasing wildlife populations and the isolation of populations through habitat 67 fragmentation (through construction of roads or other barriers that prevent animal movement) 68 69 may effectively decrease contact between humans and wildlife.

While direct human-wildlife parasite sharing is a topic of major importance, domestic animals that occur in close proximity to humans may also act as key hosts for wildlife parasites. Domestic animals (hereafter including domesticated animals, such as dogs, but also animals that live in close proximity to humans, such as commensal rats) have colonised almost all terrestrial environments (Hoberg & Brooks, 2008; Matisoo-Smith *et al.*, 1998). In turn, domestic animals commonly share subsets of their parasite fauna with humans. This subset increases the longer a species has been domesticated (Morand *et al.*, 2014; Wolfe *et al.*, 2007). Parasite host shifting at the interface between humans, domestic animals and
wildlife is a multifaceted and multidirectional problem, with potential effects for human and
wildlife health (Daszak *et al.*, 2000). Yet, while previous studies found 60 % of human
diseases to be of zoonotic origin (Taylor *et al.*, 2001; Woolhouse & Gowtage-Sequeria,
2005), global patterns in parasite sharing at the human–domestic animal–wildlife interface
are poorly resolved.

Predicting zoonotic disease risk requires understanding wildlife characteristics that 83 84 enable host shifting at local and global scales (Hoberg & Brooks, 2008). Host attributes, such as phylogenetic relatedness or overlap in habitat use, are useful for predicting whether hosts 85 share the same parasite species through ecological fitting (Streicker et al., 2010; Wells et al., 86 87 2015) or how invasions into novel environments may result in novel host-parasite associations (Agosta & Klemens, 2008; Clark et al., 2017). Conversely, knowledge of 88 whether species attributes such as demography, body size or diet increase the likelihood of 89 90 sharing parasites with humans, and whether zoonotic disease burdens in humans or domestic animals exhibit biogeographical structure, remains sparse (Han et al., 2015; Just et al., 2014; 91 Stephens et al., 2016). 92

A key gap in our understanding of zoonotic disease emergence is information on how 93 94 patterns of wildlife parasite sharing differ among domestic host species or across 95 biogeographical regions. Despite persisting in close spatial proximity, humans and domestic animals differ in habitat use, diet and other ecological traits. This may have consequences for 96 determining subsets of parasites that humans and domestic animals share with wildlife. 97 98 Humans and dogs, for example, each consume a large range of invertebrate and vertebrate species (many of which may be relevant reservoir hosts) and can access almost any type of 99 terrestrial habitat. Other domestic species, such as cows, are confined to relatively few 100

habitats and food items (e.g. grassland vegetation). One may expect that different domestic
animals will exhibit different patterns of wildlife parasite sharing and, consequently, different
potential roles as carriers of zoonotic parasites. Globally, wildlife communities occur in
distinct species assemblages according to biogeographical history, speciation events and
habitat biomes (Holt *et al.*, 2013; Kraft *et al.*, 2007; Wallace, 1876). Such biogeographical
structure may lead to spatial gradients in wildlife parasite sharing for humans and domestic
animals.

Here, we investigate possible drivers of helminth parasite (Nematoda, Cestoda, 108 109 Trematoda) sharing between wildlife and focal domestic host species (including humans) at a global scale. Using a large database of mammalian host-parasite associations, we addressed 110 two key questions: 1) Which species traits make wildlife most prone to share helminth 111 112 parasites with humans or domestic species? 2) Do patterns of wildlife parasite sharing exhibit biogeographical structure across the globe? Given that humans share parasites most 113 intensively with domestic species, we expect to find similar patterns of wildlife parasite 114 sharing among humans and domestic animals. We expect this to be especially true when 115 comparing patterns for humans and dogs, as dogs have a long domestication history and share 116 a broad range of habitats with humans. We also expect biogeographical structure in wildlife 117 assemblages to drive global patterns in wildlife-human and wildlife-domestic animal parasite 118 exchange, as different wildlife traits may facilitate or impede parasite transmission cycles and 119 120 host shifting abilities.

121

Materials and Methods

123 Host-parasite database

We compiled a global database of mammalian host–parasite associations from the publicly
available Natural History Museum (NHM), London's Host-Parasite Database (Gibson *et al.*,

126 2005). This database has been used as a backbone for the comprehensive Fauna Europaea biodiversity inventories of parasitic worms (Gibson & Bray, 1994; Gibson et al., 2014) and is 127 arguably the largest publicly available compilation of country-level records for helminth host 128 129 associations to date. In humans, for example, previous estimates suggested > 300 helminth species infecting humans (Crompton, 1999), whereas our database reports a total of 397 130 helminth species (Nematoda, Cestoda, Trematoda) to be associated with humans. We 131 downloaded all host-parasite data from the database using web-scraping tools implemented in 132 the package xml in the software R (R Development Core Team, 2017). The data of interest 133 134 for our study were country-specific combinations of parasite-mammal species associations, which included information from wild and domestic mammals as well as humans. We 135 excluded all records from captive animals or experiments, and considered only records that 136 137 included full binomial species names (scientific genus and species names). As the original database records were not specified in detail, records may include reports of molecular 138 identification of parasite species and also dead-end hosts, from which parasites are not 139 transmitted to other species. Mammal species synonyms were standardised using the 140 taxonomy of Wilson & Reeder (2005) and the IUCN Red List. Parasite names were 141 standardised using the WoRMS database (http://www.marinespecies.org), the tapeworm 142 database at the University of Connecticut (http://tapewormdb.uconn.edu/) and GBIF 143 (www.gbif.org/). Location names were standardised to country names of the current world 144 145 geopolitical map and assigned to one of 11 global zoogeographical regions according to Holt et al. (2013). Since China covers different zoogeographical regions, and not all records from 146 China could be assigned to any particular region, we classified these unspecified records into 147 148 an extra category ("China unspecified").

Our dataset consisted of 24,486 unique combinations of host–parasite–country records for
selected helminth species (Nematoda, Cestoda, Trematoda), of which 1,737 involved humans

151 as a host, from a total of 4,507 selected parasite species. Of the 1,366 total mammalian host species in our dataset, we considered 21 species as 'domestic' (including humans and 152 commensal murids) and all others as 'wildlife'. Domestic species were banteng (Bos 153 javanicus), yak (B. mutus), cow (B. taurus), bactrian camel (Camelus bactrianus), dromedary 154 (C. dromedarius), dog (Canis familiaris), goat (Capra aegagrus), guinea pig (Cavia 155 porcellus), wild ass (Equus africanus), donkey (E. asinus), horse (E. caballus), cat (Felis 156 157 catus), human (Homo sapiens), guanaco (Lama guanicoe), house mouse (Mus musculus), rabbit (Oryctolagus cuniculus), sheep (Ovis aries), brown rat (Rattus norvegicus), black rat 158 159 (R. rattus), pig (Sus scrofa) and vicugna (Vicugna vicugna). From these domestic species, we selected parasite assemblages (Nematoda, Cestoda, Trematoda) from seven focal host species 160 (hereafter termed focal species) to examine associations with wildlife: man, dog, cat, cow, 161 162 pig, black rat and brown rat. Focal host species were selected because they are among the most cosmopolitan host species and are represented with enough records in the database to 163 facilitate statistical inference of wildlife parasite sharing patterns. 164 We are aware that this dataset is incomplete in that it lacks recently described parasite 165

species and recent records of host-parasite associations in different locations; while this limits inference about absolute species numbers, we believe this dataset provides meaningful insights into the relative strength of how wildlife species share parasites with domestic species in relation to ecological traits and projected global maps, which were the main interests of this study.

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172 Host ecological traits

We selected nine ecological traits of terrestrial mammals, based on the PanTHERIA (Jones *et al.*, 2009) and EltonTraits 1.0 (Wilman *et al.*, 2014) databases, to include a broad range of
attributes likely to distinguish hosts in terms of their suitability for a parasite's life and

176 transmission cycles. Selected traits included: body mass, which is a key feature of mammals in terms of their metabolism and adaptation to environments; average longevity, litter size 177 and the average number of litters per year as demographic parameters that could be relevant 178 for allowing parasites to complete parts of their life cycles in a host; diet breadth (calculated 179 as a Shannon diversity index based on the proportional use of 10 diet categories as presented 180 in EltonTraits) and diet class ('invertebrate predator', 'herbivore', 'omnivore' or 'carnivore') 181 182 as trophic interactions traits; range area, which we expect to affect the exposure to other mammalian host species; average temperature within range as an indicator of climate niche; 183 184 and habitat as multiple binary indicators of whether a species uses 1) forest, 2) open vegetation, and/or 3) artificial/anthropogenic habitats. Information about specific habitat 185 utilisation was compiled from the International Union for the Conservation of Nature (IUCN) 186 187 database (IUCN, 2014). We did not include a larger set of ecological traits in our analysis to avoid trait autocorrelation and colinearity issues in the modelling. 188 We accounted for phylogenetic distances between wildlife species and focal domestic species 189 based on a correlation matrix (Paradis et al., 2004) of phylogenetic branch lengths, which 190 was built using a recent mammalian phylogenetic supertree (Bininda-Emonds et al., 2007). 191 We further considered the orders Carnivora, Rodentia and Primates as binary (categorical) 192 indicator variables for the major taxonomic groups that are suspected to share parasites with 193 the focal species; we used these as indicators to account for possible group-level taxonomic 194 195 effects additional to the phylogenetic branching. To account for sampling bias among wildlife species, we queried the number of published references for each binomial wildlife species 196 name from the Scopus literature database (accessed 25/02/2017); we used this measure as 197 198 more refined searches, such as the number of references linked only to parasites, included large proportions of zeros and information on the true number of sampled individuals (which 199 should determine the chance parasites are detected if prevalence is low) was not available. 200

202 Statistical analysis

203 The primary goal of this study was to identify drivers of parasite sharing between focal domestic species and wildlife. We addressed this aim using logistic hierarchical regressions 204 to analyse the relative strength of covariates that could determine the probability of parasites 205 from either humans or domestic species to be found in wildlife species (calibrated on host 206 207 species from the NHM database). For each focal domestic host species d, we constructed presence-absence vectors $Y_d(w_r, p_d)$ that encompassed all combinations of mammalian 208 209 wildlife species w_r (i.e. non-domestic species in the database) surveyed for parasites from any zoogeographical region r and all parasite species p_d from one of the selected parasite groups 210 (Nematoda, Cestoda, Trematoda) recorded in the respective focal species. Here, we assume 211 212 that any wildlife species recorded in our database has been potentially examined for all parasite species p_d known to occur in the respective region; the absence of such records are 213 set to 0. These 'absence records' likely include false zeros due to missing observations and 214 hence underestimate the link of parasite species from focal hosts to wildlife; however, we 215 prefer this approach to presence-only modelling, as the true proportion of wildlife species 216 infected remains unknown, and we thus expect techniques such as data imputation not to 217 improve our analysis. 218

We assumed the resulting data vectors $Y_d(w_r, p_d)$ are random draws from the underlying association probability $\Psi_d(w_r, p_d)$ of a wildlife species sharing a parasite with a focal species according to a Bernoulli distribution, as commonly used in logistic regression. We modelled the probability $\Psi_d(w_r, p_d)$ further using a logit-link function such that

223
$$\operatorname{logit}[\Psi_d(w_r, p_d)] = \mu_{Parasite}(p_d) + \mu_{Region}(r) + B ET_{wr}$$

where $\mu_{Parasite}(p_d)$ is the parasite-specific intercept, μ_{Region} are coefficient estimates that account for variation across zoogeographical regions *r*, and *B* is a vector of coefficient estimates that accounts for variations in the association risk linked to the matrix of covariates ET_{wr} of the nine host ecological traits, the phylogenetic distance of wildlife to the focal host species and the number of publications, as specified above.

229 We used a hierarchical model with a common hyperprior η_{μ} for the intercept as

230 $\mu(p_d) = N(\eta_{\mu}, \varepsilon_{\mu}).$

where ε_{μ} is a random Gaussian variance term that allows species-specific intercepts to vary 231 from the hyperprior (no group specific hyperprior was specified as we ran models separately 232 for the three parasite groups). We fitted the model in a Bayesian framework with Markov 233 234 Chain Monte Carlo (MCMC) sampling based on the Gibbs sampler in the freeware OpenBUGS (Lunn et al., 2012). We used a Gibbs variable selection procedure (GVS) to only 235 include variables in the model if sufficiently supported by the fit to data and joint sampling of 236 237 the most likely coefficient values (selection frequencies were typically high for covariates with significant effects, except for the categorical effects of 'region'). We used normal priors 238 with mean = 0 and variance $\sigma \sim Exp(1)$ for intercepts and all regression coefficients if selected 239 as part of the GVS, and $\sigma \rightarrow 0$ otherwise. This prior gives close approximation to a logistic 240 distribution and is appropriate for logit-scale estimates (Lunn et al., 2012). Convergence of 241 MCMC chains was assessed visually and with Gelman-Rubin diagnostics (all values < 1.2) 242 after burn-ins > 50,000 MCMC samples. Parameter estimates were calculated as posterior 243 modes and 95% highest posterior density credible intervals (CI) from 5,000 samples. 244 245 Posterior predictive checks assessed whether model assumptions were reasonable approximations of the data generating process, with Bayesian *P*-values around 0.5 indicating 246 247 a good fit. This model checking approach essentially compares whether the observed data resemble data simulated from the posterior distribution (Gelman et al., 1996). All covariates 248 249 were scaled (dividing centred values by one SD) and log-transformed if featuring overdispersion (body mass, range area, number of publications) to facilitate comparison of 250

251 effect sizes. Missing values of ecological trait covariates were imputed during MCMC updates, randomly drawing values from priors according to the mean and variance of all 252 known trait values (considering all information in the trait databases) from species in the 253 254 same orders. Specific trait data are currently not available for a considerable diversity of mammalian species; consequently, because of our ability to meaningfully impute missing 255 data using a Bayesian sampling approach, account for uncertainty in parameter estimates, and 256 257 make reasonably accurate parameter estimates, we preferred this approach over others, such as machine learning techniques, which are commonly used to more flexibly model nonlinear 258 259 and interaction relationships (Elith et al., 2008). Significance of model effects was determined by examining whether the 95% CI of regression coefficients did not overlap zero 260 for continuous covariates or were clearly distinguished from each other for categorical 261 262 covariates.

We computed the relative risk that a wildlife species will share parasites with each of 263 the focal host species for all 5,289 terrestrial mammal species in the IUCN database by 264 entering species' ecological traits into equations from the fitted models above (using posterior 265 modes of parameter estimates). We hereafter refer to this relative risk as the association risk, 266 which is appropriate in this case since the analysed data vectors included all combinations of 267 parasites from a focal host and wildlife species. Thus, the association risk would be '1' if a 268 wildlife species is likely to share all parasite species known from a particular focal species. 269 270 We set the respective parameter values to zero if trait variables were missing (i.e. assuming an 'average' effect of the respective covariate for computing the respective species-level 271 association risk). 272

The second aim of this study was to examine whether patterns of wildlife parasite sharing among domestic hosts exhibit biogeographical structure, which could be informative for understanding the future spread of zoonotic parasites. We addressed this goal by 276 exploring global patterns in observed parasite associations for focal host species and forecasted associations of wildlife infestation with parasites shared with the focal host 277 species. First, for each focal host species, we used our model outputs to generate a series of 278 maps (10km² raster cell sizes) to forecast global patterns in both wildlife parasite association 279 risks and parasite assemblage structure. Using IUCN geographical range maps for wildlife 280 species, we projected the respective parasite association risks on a global raster and, for each 281 282 cell, computed average species-level and cumulative community-level geographical association risks for local wildlife assemblages. We were not able to account for possible 283 284 regional variation in realized host-parasite interactions (which could arise due to variation in local conditions that enable host-parasite interactions) within the given wildlife range maps 285 and, for simplicity, assumed homogenous association risks throughout species' given ranges. 286

287 Next, to explore variation in parasite assemblage structure across zoogeographical regions, we computed for each cell the hypothetical presence of focal host parasites in local 288 wildlife by assuming that a parasite species occurred throughout the range of its associated 289 290 wildlife host species. We then aggregated the presence-absence of these parasites at the zoogeographical region level and calculated parasite species turnover across regions using the 291 292 β sim index, a basic turnover index that is based on the number of shared and unique species and is relatively unbiased by species richness (Lennon et al., 2001). As an index of parasite 293 assemblage distinctiveness in each region, we calculated the mean of all region-specific 294 295 pairwise β sim indices. We explored geographical sampling bias by computing the number of wildlife species examined for helminths (including species not found in domestic host 296 species) and wildlife species richness for each cell. 297

Spearman rank correlation tests were used to compare biogeographical patterns. First,
we assessed whether infestation of a greater number of focal host species leads to broader
biogeographical spread by testing the correlation strength between the Shannon index of

biogeographical spread and the total number of associated focal host species. We then
explored whether wildlife species show similar biogeographical patterns in the risk for
sharing parasites with different focal host species by testing all pairwise correlations between

the geographical association risks for parasites from the different focal host species.

We quantified the biogeographical spread of parasites (Nematoda, Cestoda, Trematoda) found in any focal species. We did this by calculating a Shannon index H_p for each parasite species p to account for both 'richness', according to the number of zoogeographical regions where a species was recorded, and 'evenness', according to the proportion $\varphi_p(r)$ of wildlife species infected with the respective parasite species in each zoogeographical region r (Magurran, 2004). We calculated the index as

311
$$H_p = \sum_{r=1}^{R} \varphi_p(r) \ln[\varphi_n(r)].$$

Larger values indicate higher proportions of wildlife species infected and a more even spreadby the parasite across zoogeographical regions.

All statistical analyses and distributional map constructions were conducted separately for the
three groups of Nematoda, Cestoda and Trematoda using R (R Development Core Team,
2017) for data preparation and summary statistics.

317

318 **Results**

Of the 1,345 mammalian wildlife species in our host-parasite database, 41 % (n = 553 spp.)

320 were infected with helminth parasite species (Nematoda, Cestoda and Trematoda) also found

in humans. For humans, in turn, 49 % (195 of 397 spp.) of all helminth parasite species were

- also found in wildlife and 45 % (182 spp.) in at least one other domestic host species. The
- 323 wildlife species associated with the highest numbers of zoonotic parasites were *Vulpes vulpes*
- 324 (red fox, 51 spp.), Canis lupus (grey wolf, 33 spp.) and Nyctereutes procyonoides (raccoon
- dog, 29 spp.). For the other focal domestic host species, proportions of examined wildlife

species that shared parasites ranged from 21 - 31 % (**Table S1, Supplementary**

327 Information).

Diet class was the strongest predictor of sharing parasites with focal host species for all 328 329 combinations of parasites (Nematoda, Cestoda, Trematoda) and focal host species, explaining 25 - 78 % of variation in wildlife infestation risk (all 95 % credible intervals, CIs = 13 - 96330 %) (Fig. 1). Wild insectivorous and omnivorous mammals were at significantly lower risk of 331 332 sharing parasites with humans than were herbivores and carnivores, a pattern that was also true for other domestic host species (with a few exceptions; Fig. S1). Risks of wildlife 333 334 species sharing parasites with the focal species also differed across zoogeographical regions. Overall, risks were relatively high in the Palaearctic region (Fig. S2), though some 335 combinations of parasite and domestic host species exhibited other informative 336 337 zoogeographical patterns. Wildlife had increased risk of sharing trematodes with cows and black rats in the Neotropical region and increased risk of sharing nematodes with humans, 338 dogs and cats in the Nearctic region. In contrast, the risk for wildlife sharing cestodes with 339 focal host species was generally low in the Neotropical region (Fig. S2). Nevertheless, the 340 overall effect of zoogeographical region was weaker than the effect of diet class (Fig. 1). 341 Coefficient estimates for all other covariates are presented in Table S2; notably, various 342 coefficient estimates were significantly different from zero, though they explained much less 343 variance than diet and zoogeographical region. Bayesian p values ranged from 0.43 to 0.79 344 345 for the various models.

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Figure 1. Relative influence (% variance explained) of wildlife host ecological traits,
zoogeographical region and number of publications (a surrogate of sampling effort) on the
probability that wildlife species shared helminth parasite species with humans or selected
domestic species. Coloured bars represent posterior modes, grey bars show 95 % credible
intervals based on the statistical sampling approach. The trait variables habitat, range area,
diet breadth and litters per years were excluded from the plot because of their negligible
effects.

347

Model-based predictions of association risk revealed two prominent patterns: first, bat species (Chiroptera) are predicted to show a low risk of sharing parasites with focal hosts (**Fig. S3**). Second, wildlife association risks were often strongly correlated across different focal hosts. The strongest of these correlations were between the risk of wildlife species sharing human cestodes and dog cestodes, human trematodes and dog trematodes, and human cestodes and dog trematodes (Spearman's r = 0.97, 0.98, and 0.96, respectively) (**Fig. S4**). 362 Wild mammals occurring in the Nearctic and Palaearctic regions were predicted to show high association risk for sharing parasites with humans, a pattern that held across all 363 three parasite groups (Fig. 2). In contrast, this predicted risk was remarkably low when 364 365 considering human-associated cestodes in the Neotropical region and human-associated parasites from all three parasite groups in the Australian region (Fig. 2). Cumulative 366 community-level association risks (summed over all wildlife species in local species pools) 367 resulted in some different patterns. The risk of sharing human parasites was high for wildlife 368 communities occurring in the Nearctic region (particularly for cestodes and trematodes) and 369 370 in mammalian diversity hotspots such as the Panamanian and Neotropical (especially for trematodes) and Afrotropical (nematodes and trematodes) regions (Fig. 2). Note that 371 relationships between observed proportions of shared parasites and the trait-based prediction 372 373 of association risks exhibited some uncertainty (Fig. S5). Nevertheless, correlations in 374 community-level association risks were even stronger than were species-level correlations, suggesting broad-scale patterns in parasite sharing are predictable (Fig. S6). We did not 375 376 identify any major global patterns in parasite assemblage distinctiveness (mean turnover in shared parasite species across zoogeographical regions), though this metric appeared to be 377 relatively higher in trematodes than in cestodes, and relatively moderate in nematodes (Fig. 378 **S7**). 379

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Figure 2. Predicted average (species-level) and cumulative (community-level) geographical
association risks for local wildlife assemblages sharing helminth parasites (Nematoda,
Cestoda, Trematoda) with humans. The risk of wildlife species sharing parasites with humans
were computed using data on host-parasite associations and ecological profiles for 1,345
wildlife species. Projections of model-based predictions on a global map are based on
computed wildlife species-level association risks for all extant mammals, rasterised at 10 km²
resolution and respective IUCN range maps.

At the parasite species level, Shannon indices describing the biogeographical spread of the 1,103 recorded helminth species followed an exponential distribution (**Fig. 3**). The most globally widespread parasite species were *Calodium hepaticum* (Nematoda),

- 393 Echinococcus granulosus (Cestoda), E. multilocularis (Cestoda), Hydatigera taeniaeformis
- 394 (Cestoda) and *Hymenolepis diminuta* (Cestoda), all of which infected 50 73 wildlife species
- and were recorded in at least three of the focal host species (**Fig. 4**). The correlation between
- the index of parasite biogeographical spread and the total number of associated focal host
- 397 species, however, was only moderate (Spearman's r = 0.5, p < 0.01).
- 398
- 399



Helminth species (ranked)



405zoogeographical regions). Colours represent the number of focal domestic host species406(human, dog, cat, cow, pig, black rat, brown rat) also associated with that parasite species.407The size of points reflects the number of zoogeographical realms in which the respective408helminth species have been recorded (1 - 12, including one class of unspecified records from409China).

410



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Figure 4. Bipartite network plot of the most globally widespread helminth parasite species associations with wildlife species pools in different zoogeographical regions. Upper nodes represent zoogeographical regions and lower nodes parasite species. The widths of links represent the relative proportion of wildlife species from the regional species pool associated with the respective parasites. Colours of lower nodes (parasites) represent the number of focal domestic host species (humans, dog, cat, cow, pig, black rat, brown rat) also associated with that parasite species, illustrating that the majority of globally spread parasites are linked to

multiple domestic host species. Widespread parasites species (n = 31) were identified as
those with the highest Shannon index scores, accounting for the associated proportions of
sampled wildlife species in different zoogeographical realms. Note the Madagascan region is
not shown, as no wildlife species were associated with the displayed parasites.

423

424 **Discussion**

425 Global biodiversity change will affect human and animal health in many ways, but potential shifts in disease burden at the human-animal interface are largely unexplored (Myers et al., 426 2013), particularly at the macro-ecological scale (Stephens *et al.*, 2016). We show that diet is 427 a key driver of the risk that wild mammal species share helminth parasites with humans. 428 Carnivores and herbivores, in particular, are at high risk of sharing parasites with humans, 429 430 while insectivores are generally at low risk. Relatively weaker effects of a wildlife species' climatic and habitat niches indicate that zoonotic parasite spread will not be contained if 431 432 contacts between wildlife and humans continue to increase. Crucially, these same patterns 433 hold when assessing the risk of wildlife sharing helminths with important domestic animals. While parasite sharing is a multifaceted one-health issue, we show that decomposing risk of 434 parasite sharing based on species' ecological and climatic niches is an important first step 435 436 towards predicting future parasite emergence.

437

438 Diet as a key driver of helminth parasite sharing

Our study focuses on terrestrial mammalian species, of which many interact in predator–prey
relationships. The completion of life cycles for some of the most globally widespread

helminths, such as *Echinococcus* spp. and *Fasciola* spp., which are also of significant health

442 concern (Garcia *et al.*, 2007), depend on such trophic interactions among mammalian hosts.

443 Unlike microparasites (viruses, bacteria, protozoa, fungi), the majority of parasitic helminth

species do not replicate in the definitive vertebrate host, with many species requiring 444 transmission through a diversity of invertebrates to complete their life cycles. For wildlife 445 insectivores, the low risks of carrying domestic animal helminths found by our study suggest 446 there is a transmission disruption that prevents host shifting (e.g. if humans consume 447 insectivorous species such as bats or shrews, but these species do not in turn ingest 448 contaminated material from humans or other infected species). Alternatively, domestic 449 450 animals and insectivorous wildlife species may not adequately share resources, such as invertebrate food items or particular habitats, which would enable parasite transmission. 451 452 The majority of parasitic nematode species undergo free-living life-history stages in the environment; some are transmitted by direct skin penetration into the definite host, whereas 453 others are transmitted through trophic interactions that may involve the ingestion of 454 455 intermediate invertebrate hosts (Anderson, 2000). This environmental transmission may play 456 an important role in governing nematode host sharing. Wild and domestic ungulate species, for example, may share considerable proportions of their nematode fauna through grazing on 457 458 common grounds (Walker & Morgan, 2014). Importantly, although one might expect hostshifting of parasites with free-living stages to be susceptible to environmental conditions, our 459 results suggest host sharing is more strongly linked to the diet strategy of the host species. 460 Focusing just on helminth parasites, we found notable differences compared to previous 461 studies examining zoonotic disease risk and reservoir potential for wildlife species. A recent 462 463 study on the zoonotic reservoir potential of rodents for both helminths and microparasites (viruses, bacteria, protozoa, fungi), for example, predicted that the rather fast-paced life 464 history strategies of rodents should be linked to a higher reservoir potential for zoonotic 465 466 diseases (Han et al., 2015). Furthermore, Luis et al. (2013) reported both bats and rodents to be major natural reservoirs for viral zoonoses. In contrast, we predicted the majority of bat 467 species are less likely to share parasites with humans and domestic species (see Fig. S4). 468

469 Different mechanisms may apply in relation to how helminths and microparasites are spread
470 through multi-species systems at the human–domestic animal–wildlife interface, warranting
471 future research.

472

480

473 Roles of domestic animal hosts at the human – wildlife interface

474 Consistent with our expectations, we found strong correlations between the risk that wildlife475 will share cestodes with human and the risk that wildlife will share cestodes with dogs.

476 Previous work has suggested that dogs and humans share a considerable number of parasites

477 (Morand *et al.*, 2014). We extend these findings to show that, concomitant with man's long

478 association with dogs and the collective exploitation of environments, both humans and dogs

479 share a considerable number of their helminth parasites with wildlife. However, this pattern is

not restricted to dogs, but can be also seen in patterns of parasite sharing for various domestic

481 species at a global scale. We found generally strong correlations in the spatially projected

482 wildlife associations risks – both at the species-level and the community-level – across

483 domestic host species (Fig. S6, S7). This emphasises, that for helminth parasites, the human-

484 wildlife interface is not independent of domestic species.

Our findings support previous calls for multi-species and community-level 485 approaches to understand parasite and disease spread (Fenton et al., 2015; Johnson et al., 486 2015; Viana et al., 2014). Notably, we provide a starting point for explaining how 487 488 overlapping distributions and contact patterns between humans, domestic animals and wildlife may impact zoonotic helminth spread at a global scale. Based on our results, future 489 geographical spread of helminth parasites will likely be facilitated through infection of 490 491 multiple domestic hosts (and possibly also invasive mammal species) that show similar trophic relationships. 492

493

We demonstrate clear zoogeographical structure in predicted risks that wildlife will

494 share parasites with humans and domestic host species. The highest risk is consistently found in the Palaearctic and Nearctic regions. Similar global patterns have been reported for rodent-495 borne zoonotic diseases, for example by Han et al. (2015). In previous work, we found the 496 497 two commensal rat species included in our study generally share helminth parasites with wildlife species of least conservation concern (Wells et al., 2015), which are likely those 498 species well adapted to anthropogenically modified landscapes. Possibly, strong adaptation to 499 anthropogenically modified landscapes by many wildlife species in the Palaearctic and 500 Nearctic regions, in combination with relevant ecological profiles, could contribute to the 501 502 strong geographical gradients in risks of parasite sharing.

503 Unfortunately, it is very difficult to discern historical host shifts by parasites, and thus 504 any possible spill-over and spill-back events, unless adequate molecular data for ancestral 505 state reconstruction are available (Hoberg *et al.*, 2001; Terefe *et al.*, 2014). Our analysis does 506 not determine whether wildlife hosts have acquired parasites from humans and domestic 507 animals, or vice versa. This is especially challenging for humans and domestic species, which 508 hardly exist in isolation from each other.

509

510 Future parasite spread through mechanisms of parasite sharing

Our finding that trophic interactions are important for interspecific helminth sharing 511 indicates the need for quantitative approaches that predict whether potential host species may 512 513 interact locally in predator-prey relationships. Our predictions can foster a better understanding of how future domestic animal and wildlife assemblages might impact 514 potential parasite host shifting through ecological fitting and changed biotic interactions (e.g. 515 516 predator-prey relationships). Zoonotic disease risk caused by helminths, for example, could then be refined to sophisticated measures that take multi-species networks of trophic 517 interactions into account, rather than only considering the number of wildlife species in local 518

assemblages (Karesh *et al.*, 2005). Given the variable sensitivities among wildlife species to
climate change, such work could also account for shifting trophic interactions among
potential parasite hosts through regionally altered community assemblages (Lurgi *et al.*,
2012).

The wildlife and domestic animal trade, together with species invasions and shifting 523 species ranges, will continue to mix formerly disjunctive host species assemblages and cause 524 biotic homogenisation (Hobbs et al., 2009). However, future climate-induced range shifts, 525 decreasing population sizes or newly arising barriers that prevent wildlife movement can also 526 527 decrease contact intensity between humans and some wildlife species. This may serve to inhibit the sharing of parasites. We nevertheless believe that very few wildlife species will be 528 sufficiently 'left alone' by humans to prevent parasite exchange unless such wildlife species 529 530 are extremely rare.

531

532 Host-parasite interactions and sampling bias

Based on records of presence-only host-parasite associations, we consider the results of our 533 study to be indicative for unravelling general patterns, rather than for providing precise 534 predictions. Several challenges are associated with studying species interactions and macro-535 ecological patterns from presence-only data. First, it is well known from sampling and 536 probability theory that parasites are likely overlooked in host species sampled with relatively 537 538 low intensity (Little, 2004). This will be especially true when low parasite prevalence prevents detection in a limited number of examined host individuals. Helminth species 539 richness in freshwater fish, for example, was found to be highly correlated with the number 540 541 of individuals examined per host species (Walther et al., 1995). An obstacle to accounting for this sampling bias is that the true sampling effort, that is the number individuals per host 542 species examined, was not available for our study; this would have enabled us to better 543

544 correct for sampling bias when making inferences about host-parasite associations (Wells et al., 2013). Moreover, spatial bias, both in the species sampled and in species-species 545 interactions, is generally known to strongly bias inference of macro-ecological patterns 546 (Boakes et al., 2010; Meyer et al., 2015). Bias may be also linked to parasite size, if large 547 species are more likely to be detected. Our trait-based approach may leverage (to some 548 extent) poorly sampled species and we used the number of publications for each host species 549 550 as a simplified proxy of sampling bias. Limitations in the currently available data on hostparasite associations and infectious disease prevent concise mapping of the majority of 551 552 parasites and diseases (Hay et al., 2013). Considering further sampling bias – as far as relevant data are available – could be of especial interest for inferring large-scale global 553 patterns. The proportion of wildlife species examined for parasites, for example, exhibits 554 555 considerable gradients across zoogeographical regions (Fig. S8). This warrants future 556 research and a critical revision of whether the particularly strong linkage of human parasites to wildlife in temperate Europe and North America, as found in this study and others (Han et 557 al., 2015; Murray et al., 2015), is a true biological phenomenon or a consequence of uneven 558 survey efforts. Moreover, improving the spatial resolution to understand whether host-559 parasite interactions and disease emergence are constrained to only those parts of a species 560 range where enabling conditions are met would improve predictions and our understanding of 561 how natural barriers may prevent disease emergence. This is of particular importance as, 562 563 ultimately, ecological and epidemiological dynamics are driving the interaction between hosts and parasites and possible parasite spill-over among hosts (Plowright et al., 2017). 564

565

Anticipating and mitigating future changes in parasite host shifting at the human– wildlife interface may require quantitative approaches that consider novel transmission pathways. These shifting pathways could be caused by the ongoing decline and/or extinction of native species (Schipper *et al.*, 2008), the introduction of invasive species (Clavero &
García-Berthou, 2005) and/or the increasing density of domestic livestock species (Jones *et al.*, 2013). Novel trophic interactions at the human–wildlife interface may also be largely
driven by human behaviour, such as expanding the menu of consumed animal species, or the
exposure of domestic species to potentially contaminated food waste (Macpherson, 2005).
Disentangling the roles of trophic and other biotic interactions versus environmental
conditions in driving parasite host sharing will improve public and wildlife health measures.

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581

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