1	Parasite prevalence increases with temperature in an avian
2	metapopulation in northern Norway
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21 Abstract

22 Climate and weather conditions may have substantial effects on the ecology of both parasites and hosts in natural populations. The strength and shape of the effects of weather on parasites and 23 24 hosts are likely to change as global warming affects local climate. These changes may in turn 25 alter fundamental elements of the parasite-host dynamic. We explored the influence of 26 temperature and precipitation on parasite prevalence in a metapopulation of avian hosts in 27 northern Norway. We also investigated if annual change in parasite prevalence was related to 28 winter climate, as described by the North Atlantic Oscillation. We found that parasite prevalence 29 increased with temperature within-years and decreased slightly with increasing precipitation. We 30 also found that a mild winter (positive winter NAO index) was associated with higher mean 31 parasite prevalence the following year. Our results indicate that both local and large scale 32 weather conditions may affect the proportion of hosts that become infected by parasites in 33 natural populations. Understanding the effect of climate and weather on parasite-host 34 relationships in natural populations is vital in order to predict the full consequence of global 35 warming.

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Key words: Climate, house sparrow, NAO, parasite, precipitation, prevalence, Syngamus
trachea, temperature

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44 Key findings

45 - Parasite prevalence increases with temperature
46 - Parasite prevalence increases following a mild winter
47 - Precipitation had little effect on parasite prevalence
48 - Climate and weather may influence parasite prevalence
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50 Introduction

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52 The relationship between parasites and climate has received increasing attention in recent years 53 due to global climate change (see Harvell et al. 2002; Altizer et al. 2013). Parasites may have a 54 significant negative effect on growth rates of host populations through their effect on fitness of 55 individual hosts (Anderson and May 1978; May and Anderson 1978). Climate change may alter 56 important aspects in the parasite-host relationship (e.g. transmission rate) and potentially lead to 57 drastic changes in the population dynamics of both parasites and hosts (Harvell et al. 2002; 58 Altizer et al. 2013). One important parameter that may be altered is the percentage of hosts 59 infected in a population at a given time, i.e. parasite prevalence. Several experimental and 60 observational studies have indicated that variation in climatic parameters may affect the parasite 61 prevalence in a population (see Harvell et al. 2009; Lafferty 2009), both direct and indirectly. For example, Zamora-Vilchis et al. (2012) found that prevalence of blood parasites was higher in 62 63 bird populations that inhabited areas with higher annual temperatures. They attributed this 64 pattern to both higher vector abundance and faster development of the parasites at higher 65 temperatures. Along with temperature, there has also been a call for more studies which also investigate the effects of precipitation (Altizer et al. 2013). Precipitation may have a direct 66

67 influence on parasite prevalence, for example through desiccation of free-living parasite eggs or 68 larvae (e.g. Hudson and Dobson 1991; Miterpakova et al. 2006; Jex et al. 2007) or indirectly 69 through other mechanisms such as vector availability (Sacks et al. 2004). Furthermore, host 70 demography may play an important role in the parasite-host relationship (see Dobson 1990). For 71 instance, transmission of parasites may increase with host population density (Anderson and 72 May 1978), which in turn may increase prevalence (Arneberg 2001). Parasites often have an 73 aggregated distribution with respect to host individuals (see Wilson et al. 2002). Therefore, 74 reliable estimates of parasite prevalence require sampling of a large proportion of the host 75 population (Wilson et al. 2002). In addition, because parasite prevalence may differ between 76 different classes of individuals (e.g. adults vs. juveniles), sampling should be stratified (i.e. equal 77 proportion of different groups) in order to avoid bias in estimates of prevalence. 78 In the present study we explore the relationship between climate and prevalence of the 79 parasitic nematode Syngamus trachea in a metapopulation of house sparrows (Passer domesticus) in northern Norway over a 7-year period. Our aim was to investigate whether 80 81 within-year variation in prevalence of Syngamus trachea could be explained by temperature, 82 precipitation and population density. We also investigated the relationship between prevalence 83 among years and the preceding winter climate estimated by the North Atlantic Oscillation index

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85

86 Material and methods

87 Study area

(NAO).

88 The study was carried out in an archipelago consisting of 18 islands in the Helgeland district in

89 northern Norway (Supplementary Fig. S1). The house sparrows on these islands have been

90 studied on an individual based level by capture-mark-recapture methods since 1993 (Holand et 91 al. 2016). In this study we focus on 4 of the islands (Aldra, Gjerøy, Hestmannøy and Indre 92 Kvarøv), where the sparrows mainly live in colonies on dairy farms. House sparrows shelter 93 inside farm buildings when the weather is harsh, especially in winter. A high proportion of 94 individuals have been individually marked (> 90%), the annual recapture probability is high (> 95 70 %), and fecal samples have been collected since 2007. Individuals have often been captured 96 and sampled several times during their lifetime. Because of the sedentary nature of the house 97 sparrows, the large size of the study area and the high proportion of individually marked 98 individuals in our metapopulation, we can identify individuals that move between islands in the 99 study area (Pärn et al. 2012).

100

101 Study species

102 The parasite of interest in this study is the nematode worm Syngamus trachea (henceforth S. 103 trachea). S. trachea has a cosmopolitan distribution and has been found in most terrestrial bird 104 genera (Atkinson et al. 2008). In Norway the parasite has been recorded in several wild bird 105 species (for list see Bakke 1973). The parasite is known as "gapeworm" in the poultry industry, 106 where outbreaks have caused substantial problems (Atkinson et al. 2008). S. trachea has a 107 lifecycle that may include a paratenic host (earthworms, snails or insects) in addition to the final 108 avian host. The parasite(s) feed on blood in the trachea tissue causing mechanical damage, 109 anaemia, inflammation and excess mucus production. The birds that are infected may develop 110 symptoms consisting of gasping, wheezing and shaking of the head. This discomfort may affect 111 food uptake, which combined with blood loss and the symptoms listed above, can cause death 112 (Atkinson et al. 2008). The prepatent period (i.e. the time between infection and first egg

113	produced by the parasite) is 12-17 days in chickens (Gallus gallus) and the adults can live 2-4
114	months (Barus 1966b). Males and females join in permanent copulation and the female releases
115	eggs which are passed in the feces of the host (Atkinson et al. 2008). The development time of
116	embryos into third stage invasive larvae depend on temperature, and has a lower limit of 16 °C
117	(Barus 1966a). For temperatures over 16 °C the development time may vary from ca. 45 days at
118	17 °C to ca. 7 days at 29 °C (Barus 1966a). At temperatures above 29 °C the development time
119	increases, becomes irregular and stops at 35-37 °C (Barus 1966a). Barus (1966a) found that eggs
120	could be vulnerable to desiccation and may become incapable of further development after 14
121	days of exposure to the daily varying temperature (12-24 °C) and humidity (36-97 %). It was also
122	found that c. 35 % of eggs died when exposed to temperatures between -1525 °C for 45 days.
123	The house sparrow is a small passerine bird with a global distribution, and is widely
124	distributed on the Northern hemisphere (Anderson 2006). The house sparrows live in close
125	association with human settlements. During winter their diet mainly consist of cattle food, and
126	seeds from cultivated crops (Anderson 2006). In the summer the house sparrows forage
127	outdoors, to a large extent, on different types of seeds and insects (Anderson 2006).
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129	Field work, climate data and population size
130	We captured house sparrows using mist-nets and placed the birds in paper bags with a small hole
131	at the top to prevent asphyxiation. After approximately 10 minutes in the paper bag, a fecal
132	sample had usually been obtained. Samples were stored in 1 ml of MilliQ H2O in a 1.5 ml cryo-
133	tube at between 1 and 8 °C until analyzed. Samples were analyzed by flotation of eggs in a
134	McMaster egg counting chamber (see Holand et al. 2013 for detailed descriptions). During 2007
135	– 2013 (1 st May – 8 th November), 3619 fecal samples were collected from 1705 individuals.

Detection of eggs from *S. trachea* in bird feces is a reliable and consistent method of monitoring prevalence in host populations (Barus 1966b; Barus 1966c). To our knowledge, *S. trachea* does not present a circadian rhythm in the release of eggs. In addition to captures, we also visited nests in the populations and marked fledglings when they were c. 11 days old.

140 Weather data was obtained from Norwegian Meteorological Institute's weather stations 141 on the islands Myken and Solvær in the Helgeland archipelago (Supplementary Fig. S1). High 142 spatial correlation in daily mean temperature (°C) and daily precipitation (mm) has been found 143 between the two weather stations (Ringsby et al. 2002). When choosing our weather predictor 144 variables, the life-history of the parasite was considered. The time between infection and first 145 egg produced by the parasite (prepatent period) is c. 15 days (see above). During this period we 146 assumed that weather had little or no impact on the parasite. Furthermore, we assumed that the 147 mean weather conditions during a 20-day time-window before the prepatent period was likely to 148 capture the important period during which the weather conditions (temperature and precipitation) 149 may influence the development time of infective larvae (see above). This would allow us to 150 investigate if variation in the mean weather conditions preceding our sample explained the 151 individual probability of infection by S. trachea. As an index of large-scale climatic variation we 152 used station based Hurrell North Atlantic Oscillation index (NAO) for winter months (December 153 – March). The data was obtained from the National Center for Atmospheric Research, USA. This 154 index is based on the difference in normalized atmospheric pressure at sea level between Lisbon, 155 Portugal and Reykjavik, Iceland. In Scandinavia, a positive winter NAO index is correlated with 156 milder and wetter winters whereas a negative NAO index is correlated with colder and dryer 157 winters (Hurrell 1995).

158	An estimate of population size on a given day of the year was obtained from estimates of
159	the number of alive adults and simulations of a model of daily juvenile mortality rates as a
160	function of fledging date of nestlings(for detailed method, see Supplementary information S1).
161	Relative daily population size within each sub-population was calculated by subtracting the mean
162	population size and dividing by the standard deviation (i.e. a z-score transformation). Assuming
163	that the size of the suitable habitat in the subpopulations did not change in the period, this
164	estimate corresponds to population density at a given day and subpopulation. We also calculated
165	the mean population density in a 20-day time-window before the prepatent period as we assumed
166	that this was likely to influence parasite prevalence (same as weather variables, see above)
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169	Statistical analyses
170	All analyses were done using the statistical software package R version 3.5.1 (R Core Team
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171 172 173 174	2018). The focal response variable in the statistical analyses was individual infection status. We defined an individual as infected (status = 1) if at least one egg of <i>S. trachea</i> were found in the fecal sample. If no eggs were found, the infection status was set to 0. We modeled infection status (prevalence) using generalized linear mixed models (package glmmTMB version 0.2.2.0)
171 172 173 174 175	2018). The focal response variable in the statistical analyses was individual infection status. We defined an individual as infected (status = 1) if at least one egg of <i>S. trachea</i> were found in the fecal sample. If no eggs were found, the infection status was set to 0. We modeled infection status (prevalence) using generalized linear mixed models (package glmmTMB version 0.2.2.0) with a Bernoulli response variable (i.e. whether an individual was infected or not) and a logit link
 171 172 173 174 175 176 	2018). The focal response variable in the statistical analyses was individual infection status. We defined an individual as infected (status = 1) if at least one egg of <i>S. trachea</i> were found in the fecal sample. If no eggs were found, the infection status was set to 0. We modeled infection status (prevalence) using generalized linear mixed models (package glmmTMB version 0.2.2.0) with a Bernoulli response variable (i.e. whether an individual was infected or not) and a logit link function. Because many individuals (N = 774) were sampled more than once within a season, we
 171 172 173 174 175 176 177 	2018). The focal response variable in the statistical analyses was individual infection status. We defined an individual as infected (status = 1) if at least one egg of <i>S. trachea</i> were found in the fecal sample. If no eggs were found, the infection status was set to 0. We modeled infection status (prevalence) using generalized linear mixed models (package glmmTMB version 0.2.2.0) with a Bernoulli response variable (i.e. whether an individual was infected or not) and a logit link function. Because many individuals (N = 774) were sampled more than once within a season, we included a random intercept for individual identity to avoid pseudoreplication. We have
 171 172 173 174 175 176 177 178 	2018). The focal response variable in the statistical analyses was individual infection status. We defined an individual as infected (status = 1) if at least one egg of <i>S. trachea</i> were found in the fecal sample. If no eggs were found, the infection status was set to 0. We modeled infection status (prevalence) using generalized linear mixed models (package glmmTMB version 0.2.2.0) with a Bernoulli response variable (i.e. whether an individual was infected or not) and a logit link function. Because many individuals (N = 774) were sampled more than once within a season, we included a random intercept for individual identity to avoid pseudoreplication. We have previously shown (Holand <i>et al.</i> 2013) that prevalence varies within season, islands and among

181 the date when a bird was sampled, where day 1 was 1st of January. In addition, Holand et al. 182 (2013) found substantial variation in prevalence among years. A random intercept for year was 183 therefore included in all models. To test if the effects of weather on prevalence varied during the 184 season, we included the two-way interactions between weather variables and SDN. We also 185 included second order terms of the weather variables and the population density variable in order 186 to evaluate nonlinear effects of these respective variables. No higher-order terms (interactions or 187 quadratic terms) were included without the presence of the respective main effects of the term. 188 The NAO index was included as a yearly variable (i.e. one estimate for each year in the study). 189 To increase interpretability of interaction terms, we centered SDN and the two weather variables 190 on zero (i.e. mean = 0). In order to prevent possible effect of multicollinearity, all explanatory 191 variables were assessed for collinearity visually, by the use of Pearson's correlation coefficient 192 (r_p) and variance inflation factors (VIF, see Dormann *et al.* 2013). Candidate models were ranked 193 using Akaike information criterion corrected for small sample sizes (AIC_c), which penalizes 194 models with a high number of parameters relative to sample size (Burnham and Anderson 2002). 195 Estimates from models are reported as mean ± 1 standard error (SE). 196 We also tested the correlation between observed annual mean prevalence in July and the

we also tested the conclusion between observed annual mean prevalence in Jury and the winter NAO index (n = 7). July was the month with largest sampling size of feces in our study system (Holand *et al.* 2013).

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200 Results

Six models obtained a $\Delta AIC_C < 2$ in our model comparison (Table 1). Among these, the second highest ranked model contained the fewest number of parameters and may be considered the most parsimonious. The five alternative models contained some of the parameters that were

204	present in the second highest ranked model (Table 1). All of our highest ranked models
205	contained a quadratic effect of temperature on parasite prevalence (Fig. 1 A), which indicated
206	that parasite prevalence increased to a peak at 12.37 degrees Celsius and then leveled out at very
207	high temperatures (see Fig. 1 A and Table 1). All the highest ranked models also contained a
208	negative effect of precipitation on parasite prevalence, (see Fig. 1 B and Table 1). The five
209	alternative models also contained interactions between SDN and temperature and precipitation
210	respectively, but the uncertainty of parameter estimates was large and impeded any
211	interpretations. All of the six top models contained a positive effect of NAO on parasite
212	prevalence (Fig. 1 C and Table 1). Finally, there was a strong positive correlation between mean
213	parasite prevalence in July and NAO ($r_p = 0.83$, $n = 7$, $t = 3.38$, $P = 0.02$, Fig. 2).
214	
215	Discussion
216	This study has found that climate and weather can be important factors in explaining variation in
217	parasite prevalence patterns in natural populations of hosts. Our study also indicated that large

218 scale climatic patterns measured as NAO, may have a substantial impact of parasite prevalence

219 in wild populations of hosts.

Specifically, we found that within-year prevalence of *S. trachea* was positively associated with temperature (but less so at very high temperatures). This supports previous qualitative studies conducted by Barus (1966), who observed that prevalence of *S. trachea* in domestic birds increased during the year as the temperature increased. Similar increases in prevalence within years have been found for other parasite species and hosts (e.g. Karvonen *et al.* 2010; Paredes-Esquivel *et al.* 2012; Bottari *et al.* 2013). The underlying mechanisms for a positive relationship between parasite prevalence and within-year temperature may be difficult to determine.

227 However, studies have shown that thresholds in temperature may make it very difficult for 228 certain parasites to complete their life cycles due to reduced larvae survival probability or 229 prolonged egg development time at certain temperatures (e.g. Barus 1966a; O'Connor et al. 230 2006; Tinsley et al. 2011). This may typically limit transmission of parasites to the summer 231 season in artic regions (Kutz et al. 2005; Jenkins et al. 2006; Hoberg et al. 2008). Several studies 232 have also shown that development rates of parasite eggs and larvae increase with temperature 233 (Hoar et al. 2012; Leathwick 2013). This may increase transmission rate at higher temperatures 234 (Kutz et al. 2005) and may lead to higher parasite prevalence. 235 Our results suggested that high levels of precipitation reduced parasite prevalence. 236 However, we found only minor changes in predicted prevalence for each mm of increased 237 precipitation (Fig.1 B). Thus, although eggs and larvae of S. trachea may generally be 238 susceptible to desiccation (see Material and Methods), this does not seem to a limiting factor in 239 our study system. We found no effect of population density on prevalence of S. trachea in our 240 subpopulations. This was somewhat surprising given that studies often suggest that prevalence 241 should increase with increasing population density (see Introduction). However, there may be a 242 more complex relationship between population density and transmission rates in our populations. 243 For example, parasite prevalence and transmission rates could be affected by heterogeneity in 244 parasite resistance of individual hosts (Borg et al. 2011, Lundregan et al. unpublished results). 245 Alternatively, one or more of our assumptions used to estimate population density may have 246 caused a weakening of the relationship between parasite prevalence and the true population 247 density.

Interestingly, all the highest ranked models included the annual winter NAO index, where
mild winters were associated with higher prevalence the following sampling period (1 May – 8

250 November). This result was also shown as a very strong positive correlation between winter 251 NAO index and observed mean prevalence in July the following summer (Fig. 2). However, 252 these results were based on only 7 NAO values (7 years) and should therefore be interpreted with 253 some caution. Given that this relationship is valid, it may be generated via several mechanisms. 254 First, it is known that many free-living stages of parasites (including *S. trachea*) have a relatively 255 low tolerance to temperatures below 0 °C and may suffer an increased mortality due to low 256 temperatures during winter (Barus 1966a; O'Connor et al. 2006; Descamps 2013; Sherrard-Smith 257 et al. 2013). A mild winter may result in increased winter survival of eggs and larvae, which in 258 turn may increase the parasite prevalence in the host population the following summer. 259 Alternatively, mild winters may result in increased survival probability of host individuals with a 260 poor body condition that may be more susceptible to infection the following summer. Second, a 261 positive winter NAO index is often associated with milder spring temperatures (Helama and 262 Holopainen 2012) and could therefore result in a longer transmission window for the parasite. 263 Finally, a positive winter NAO index might lead to a higher abundance of paratenic hosts 264 (insects) the following summer (Westgarth-Smith et al. 2012), thereby facilitating the 265 transmission of S. trachea to its avian hosts. Some climate models have predicted that 266 frequencies of positive winter NAO index's may increase in the future (Kuzmina et al. 2005). 267 This could lead to significant changes in the parasite-host relationship in the future. For instance, 268 as global warming is predicted to result in wetter and milder winter temperatures in northern 269 Europe (Alcarro et al. 2007), the present study suggests that parasite species which are limited 270 by low temperatures during the winter season may extend their geographical distribution to 271 higher altitudes and latitudes as a consequence. Understanding how climate impacts parasites 272 and hosts may be vital in accurately predicting the changes in natural communities due to climate

274	management efforts of threatened species and ecosystems (Harvell et al. 2002).
275	
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436 Tables

Table 1: Parameter estimates (\pm 1 SE) of the six best models (according to AIC_C) of parasite prevalence (Syngamus trachea) in a metapopulation of house sparrows in Norway. Models were fitted using generalized linear mixed models. Estimates are shown for each variable contained in the model. In addition, all models also contained the main fixed effect of island, life stage group (adult male, adult female, juvenile) and random intercept effects of individual identity and year. T = temperature, P = precipitation, SDN = season day number, df = degrees of freedom. All estimates are given with \pm 1 SE.

Model	Intercept	SDN	NAO	Т	T^2	Р	\mathbf{P}^2	T:SDN	P:SDN	df	ΔAIC_C
1	-7.410 ± 0.508	0.0004 ± 0.0033	0.256 ± 0.050	0.454 ± 0.069	-0.085 ± 0.019	-0.086 ± 0.037			0.001 ± 0.001	14	0
2	-7.429 ± 0.508	0.0008 ± 0.0032	0.269 ± 0.049	0.477 ± 0.067	-0.084 ± 0.019	-0.047 ± 0.024				13	0.08
3	-7.352 ± 0.507	0.0003 ± 0.0033	0.268 ± 0.051	0.468 ± 0.070	-0.084 ± 0.019	-0.052 ± 0.047	-0.004 ± 0.004		0.001 ± 0.001	15	0.73
4	-7.432 ± 0.509	0.0004 ± 0.0033	0.241 ± 0.052	0.429 ± 0.073	-0.069 ± 0.025	-0.081 ± 0.037		-0.002 ± 0.002	0.001 ± 0.001	15	1.04
5	-7.445 ± 0.509	0.0007 ± 0.0032	0.259 ± 0.051	0.461 ± 0.071	-0.073 ± 0.025	-0.040 ± 0.026		-0.001 ± 0.002		14	1.56
6	-7.375 ± 0.508	0.0003 ± 0.0033	0.253 ± 0.053	0.442 ± 0.074	-0.067 ± 0.025	$\textbf{-0.048} \pm 0.047$	-0.004 ± 0.004	-0.001 ± 0.002	0.001 ± 0.001	16	1.74

444

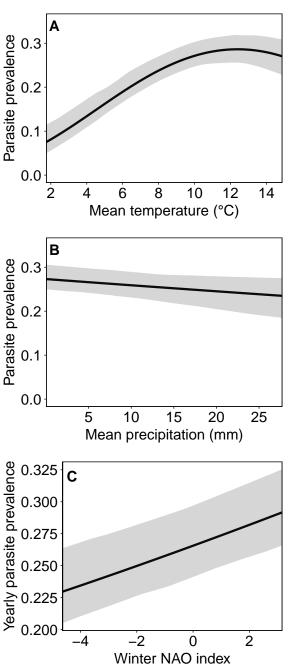
445 **Figure legends**

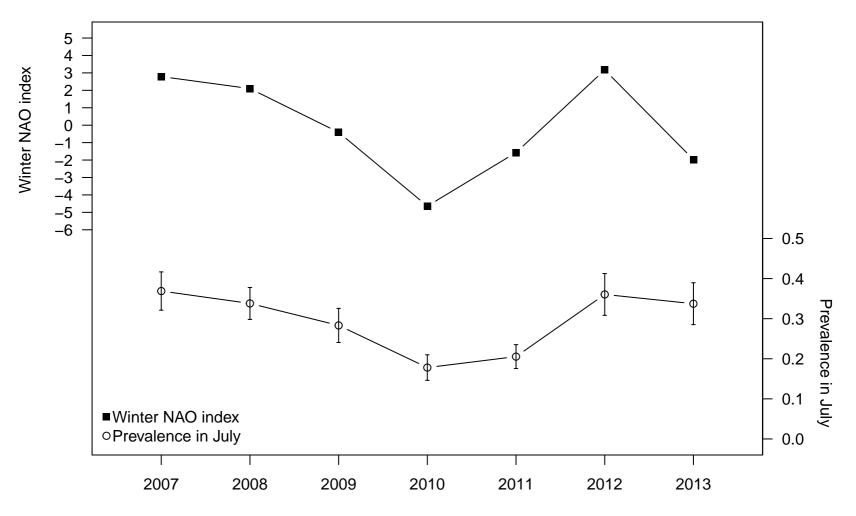
446

447	Fig.1 Predicted values (solid line) from the most parsimonious model (based on AICc) for
448	prevalence of the parasite Syngamus trachea in a metapopulation of house sparrows on the coast
449	of Helgeland, Norway. (A) The effect of within-year variation in temperature, (B) precipitation,
450	and (C) winter NAO index. All predicted values are shown with 95 % confidence intervals

451 (grey).

- 452 Fig. 2: The relationship between winter NAO index and observed mean prevalence of the
- 453 parasite *Syngamus trachea* in a metapopulation of house sparrows on the coast of Helgeland,
- 454 Norway, in the month of July. Estimates of mean July prevalence are shown with ± 1 standard
 455 error.





Year