

1 **Acclimation to warmer temperatures can protect host populations from**  
2 **both further heat stress and the potential invasion of pathogens**

3 Tobias E. Hector<sup>1,3,\*</sup>, Marta S. Shocket<sup>2</sup>, Carla M. Sgrò<sup>1</sup> & Matthew D. Hall<sup>1#</sup>

4 <sup>1</sup>School of Biological Sciences, Monash University, Melbourne, Victoria 3800, Australia

5 <sup>2</sup>Department of Ecology and Evolutionary Biology, University of California, Los Angeles, CA, USA

6 \* Current address: Department of Zoology, University of Oxford, Oxford, United Kingdom

7 # corresponding author, [matthew.hall@monash.edu](mailto:matthew.hall@monash.edu)

## 8 Abstract

9 Thermal acclimation can provide an essential buffer against heat stress for host populations,  
10 whilst acting simultaneously on various life-history traits that determine population growth. In  
11 turn, the ability of a pathogen to invade a host population is intimately linked to these changes via  
12 the supply of new susceptible hosts, as well as the impact of warming on its immediate infection  
13 dynamics. Acclimation therefore has consequences for hosts and pathogens that extend beyond  
14 simply coping with heat stress – governing both population growth trajectories and, as a result, an  
15 inherent propensity for a disease outbreak to occur. The impact of thermal acclimation on heat  
16 tolerances, however, is rarely considered simultaneously with metrics of both host and pathogen  
17 population growth, and ultimately fitness. Using the host *Daphnia magna* and its bacterial  
18 pathogen, we investigated how thermal acclimation impacts host and pathogen performance at  
19 both the individual and population scales. We first tested the effect of maternal and direct  
20 thermal acclimation on the life-history traits of infected and uninfected individuals, such as heat  
21 tolerance, fecundity, and lifespan, as well as pathogen infection success and spore production. We  
22 then predicted the effects of each acclimation treatment on rates of host and pathogen  
23 population increase by deriving a host's intrinsic growth rate ( $r_m$ ) and a pathogen's basic  
24 reproductive number ( $R_0$ ). We found that direct acclimation to warming enhanced a host's heat  
25 tolerance and rate of population growth, despite a decline in life-history traits such as lifetime  
26 fecundity and lifespan. In contrast, pathogen performance was consistently worse under warming,  
27 with within-host pathogen success, and ultimately the potential for disease spread, severely  
28 hampered at higher temperatures. Our results suggest that hosts could benefit more from  
29 warming than their pathogens, but only by linking multiple individual traits to population  
30 processes can the full impact of higher temperatures on host and pathogen population dynamics  
31 be realised.

32

33 Keywords: thermal limits, host-pathogen interactions, heat stress, *Pasteuria ramosa*, aquatic  
34 ectotherm, population growth, fitness, virulence, knockdown times

## 35    **Introduction**

36    Processes at every level of biological organisation are fundamentally shaped by temperature, from  
37    the rate at which physiological processes occur within the body, to the growth and persistence of  
38    a population or community (Angilletta et al., 2004; Chown et al., 2010; Colinet et al., 2014;  
39    Somero, 2010; Vázquez et al., 2015). This is particularly true for host-pathogen interactions  
40    (Thomas & Blanford, 2003). For a host, acclimation to rising temperatures can increase heat  
41    tolerance, providing individuals with a crucial buffer against future heat stress (Rohr et al., 2018;  
42    Sgrò et al., 2016; Sinclair et al., 2016). Warmer temperatures will also typically accelerate the pace  
43    of life, leading to earlier reproductive output and shortened lifespans, and changes to population  
44    growth rates as a result (Angilletta et al., 2004; Debecker & Stoks, 2019; Hector et al., 2021). In  
45    turn, for a pathogen, warming can accelerate the infection process by increasing within-host  
46    replication and virulence (Fels & Kaltz, 2006; S. E. Mitchell et al., 2005; Paull et al., 2015; Vale et  
47    al., 2008), thereby influencing the rates at which hosts can be encountered and infected (Kirk et  
48    al., 2018; Shocket, Strauss, et al., 2018; Shocket, Vergara, et al., 2018), and even reduce the  
49    capacity of hosts to respond to both average temperature shifts and extreme heat (Gehman et al.,  
50    2018; Greenspan et al., 2017; Hector et al., 2019; Kunze et al., 2022; Porras et al., 2021; Ware-  
51    Gilmore et al., 2021).

52    The response of hosts and pathogens to increasing temperatures will thus depend on the relative  
53    effects of thermal acclimation on thermal tolerances, versus on traits that underlie whether a host  
54    or pathogen population will increase or decrease under warming (Gehman et al., 2018; Mordecai  
55    et al., 2019; Shocket, Ryan, et al., 2018). An understanding of both the individual and population  
56    level performance begins by comparing changes in thermal tolerances (e.g., upper thermal limits  
57    or immobilisation times, (Hector et al., 2019, 2020)) with host and pathogen vital metrics. For  
58    hosts, life tables can be used to estimate the influence of thermal change on the intrinsic rate of  
59    population growth, denoted as  $r_m$ , and therefore the temperature which is likely to maximise host  
60    fitness (Amarasekare & Coutinho, 2013; Amarasekare & Savage, 2012). In contrast, the ability of a  
61    pathogen to persist in a host population can be influenced by both host population dynamics and  
62    pathogen transmission (Aulsebrook et al., 2023; Civitello et al., 2013; S. R. Hall et al., 2009;  
63    Mordecai et al., 2019). Estimates of infection success, proliferation, and the supply of new  
64    susceptible hosts can be integrated via an epidemiological model into the basic reproduction  
65    number,  $R_0$ , which captures the potential of a pathogen to spread through a completely

66 susceptible host population (Anderson & May, 1986), and its dependence on changing  
67 temperatures (Mordecai et al., 2019; Shocket, Ryan, et al., 2018; Shocket, Strauss, et al., 2018).  
68 Despite the importance of thermal tolerances in shaping host and pathogen geographic  
69 distributions (e.g., (Blanford et al., 2013; Mordecai et al., 2019; Rohr & Raffel, 2010; Shocket,  
70 Ryan, et al., 2018)), the impact of thermal acclimation on heat tolerances is rarely considered in  
71 unison with metrics of how well both host and pathogen populations might perform, such as  $r_m$   
72 and  $R_0$  (but for a non-disease example, see (Cavieres et al., 2020)). More commonly, changes in  
73 thermal tolerances are correlated with the variation induced in an individual's phenotype, such as  
74 how temperature impacts host development, fecundity, survival, or immunity, or how  
75 temperature alters pathogen proliferation or virulence (e.g., (Fels & Kaltz, 2006; Hector et al.,  
76 2019; Laidlaw et al., 2020; S. E. Mitchell et al., 2005; Paull et al., 2015; Raffel et al., 2006, 2013,  
77 2015; Sun et al., 2022; Vale et al., 2008)). Although population-level processes intrinsically depend  
78 on these individual responses to temperature, one does not necessarily predict the other (M. D.  
79 Hall & Mideo, 2018; Mideo et al., 2008; Penczykowski et al., 2016). Even where one or more  
80 measures of host or population growth are available, understanding the potential spread of the  
81 pathogen is often the primary goal and more emphasis is thus placed on  $R_0$  or related metrics  
82 (Beck-Johnson et al., 2017; Kirk et al., 2018; Mordecai et al., 2019; Paaijmans et al., 2009; Shocket  
83 et al., 2019; Shocket, Vergara, et al., 2018); but see (Gehman et al., 2018; Shocket, Ryan, et al.,  
84 2018)), rather than an explicit comparison of how warming shapes host and pathogen population  
85 growth rates and the temperature most likely to maximise fitness in each (sensu (Amarasekare &  
86 Coutinho, 2013; Amarasekare & Savage, 2012)).

87 To add complexity to our understanding of thermal change and host-pathogen interactions, an  
88 individual's prior thermal exposure can have a greater influence on fitness than contemporary  
89 temperatures. In many species, maternal and developmental acclimation are a vital mechanism  
90 for preparing populations to cope with future environmental conditions (Beaman et al., 2016;  
91 Hoffmann et al., 2012; Sgrò et al., 2016). In *Drosophila melanogaster*, for example, the  
92 temperature experienced during development can have an overriding effect on adult heat  
93 tolerance (Kellermann et al., 2017; Slotsbo et al., 2016). Temperature can also impact disease-  
94 related traits across generations and infection cycles. Host resistance to infection and the  
95 infectivity of pathogen spores have both been shown to increase when past generations  
96 experienced warmer conditions (Ferguson & Sinclair, 2019; Garbutt et al., 2014; Paull et al., 2015;  
97 Shocket, Vergara, et al., 2018; Sun et al., 2022). It remains unclear, however, how various host and

pathogen fitness components, and therefore population level dynamics, will be shaped by the thermal acclimation of a host before it encounters a pathogen.

In this study, we contrasted how thermal acclimation, before and during infection, shapes both host heat tolerance and the resulting life-history of the host and pathogen. We then expanded our view to consider host population growth and the potential spread of a pathogen. To address these questions, we used the water flea *Daphnia magna* and its bacterial pathogen *Pasteuria ramosa*. *Daphnia* are able to mount strong plastic thermal acclimation responses in their heat tolerance (Burton et al., 2020; Yampolsky et al., 2014), and temperature can mediate their response to infection (Auld & Brand, 2017; Garbutt et al., 2014; Kirk et al., 2018; Kunze et al., 2022; Shocket, Vergara, et al., 2018), and any damage that may follow (Hector et al., 2021). *Pasteuria ramosa* is a natural bacterial pathogen of *Daphnia*, which has distinct genotypes known to vary in various aspects of within-host performance, including infection rates and spore production (Clerc et al., 2015; M. D. Hall & Mideo, 2018), both of which have been shown to be sensitive to temperature stress (Vale et al., 2008; Vale & Little, 2009).

We considered two forms of thermal acclimation (at 25°C) versus a standard temperature (at 20°C): i) a maternal and developmental acclimation treatment that occurred prior to infection, and ii) a direct thermal acclimation treatment that was applied to focal individuals from the infection period onwards. Under combinations of both acclimation treatments (herein maternal and focal), we measured individual-level traits including host heat tolerance (assessed as knockdown time under heat shock, see (Hector et al., 2019), host lifespan, and the fecundity of infected and uninfected *Daphnia*, as well as within-host pathogen spore loads and infection success for each pathogen genotype. From our experiment we then used host life table data (lifespan and fecundity) to evaluate potential host population dynamics, and then parameterised an epidemiological model to estimate a metric for the potential for disease spread through a host population (Aulsebrook et al., 2023; Civitello et al., 2013; Shocket, Vergara, et al., 2018). Together, these measures allow us to contrast how thermal acclimation can impact host and pathogen performance at an individual scale, and, in turn, how they may drive population dynamics.

## Methods

### Host and pathogen

The cyclically parthenogenic crustacean *Daphnia magna* Straus is commonly found in both fresh and brackish waters, including shallow pools and large lakes, across Eurasia. *Pasteuria ramosa*

129 Metchnikoff is a Gram-positive bacterial pathogen of *D. magna* that enters the host during filter  
130 feeding, before severely reducing host fecundity (via castration) and lifespan (Clerc et al., 2015;  
131 Ebert et al., 2016; M. D. Hall et al., 2019). At host death millions of spores are released into the  
132 environment where exclusively horizontal transmission takes place, which itself depends on the  
133 interplay between the pathogen's ability to produce mature transmission spores and its virulence  
134 (M. D. Hall & Mideo, 2018). In this study we used *Daphnia* genotype BE-OMZ-M10 infected with  
135 one of three *P. ramosa* genotypes (C1, C14 and C20). These pathogen genotypes were chosen  
136 because they display genetic variation in their virulence and transmission potential (Clerc et al.,  
137 2015; M. D. Hall & Mideo, 2018), and in the extent to which they reduce host heat tolerances  
138 (Hector et al., 2019).

139 Before the experiments, female *Daphnia* taken from stock culture were placed individually in 70-  
140 mL jars filled with 50 mL of Artificial *Daphnia* Medium (ADaM; following (Ebert et al., 1998)) for  
141 three generations to minimise trans-generational effects. *Daphnia* were changed into fresh ADaM  
142 twice a week and fed with algae (*Scenedesmus sp.*) daily. To meet the growing energy needs of the  
143 animals, food levels were increased from one million cells per jar at birth, to eight million by age  
144 14 days. *Daphnia* were maintained under standard conditions (20°C, 16L:8D) and repositioned  
145 within the incubator regularly to minimise any positional effects.

#### 146 **Experimental animals, thermal acclimation, and infection**

147 Thermal acclimation began in the maternal generation. On the day of birth, maternal generation  
148 (F0) *Daphnia* were taken from clutches 3–5 of the standardised animals and maintained  
149 individually at either 20°C or 25°C (maternal/developmental temperature treatment, hereafter  
150 maternal acclimation). Experimental (F1) *Daphnia* were then collected from clutches 3–5 of the  
151 acclimated mothers on the day of birth and placed at either 20°C or 25°C (focal acclimation) in a  
152 fully factorial design, resulting in four thermal acclimation treatments (20-20, 20-25, 25-20 & 25-  
153 25°C). The maternal acclimation involved maternal and developmental effects (because *Daphnia*  
154 are ovoviviparous), whilst the focal acclimation was experienced directly by experimental animals.  
155 Experimental animals were kept at their focal acclimation temperatures from birth until either  
156 being used in heat tolerance assays or until death, including over the infection period. The warm  
157 acclimation temperature was chosen as it is ecologically realistic for the higher temperatures  
158 experienced in summer in *Daphnia* populations (Yampolsky et al., 2014) and below the thermal  
159 maxima for both the host and pathogen (Hector et al., 2019; Kirk et al., 2018).

160 The experimental generation included a total of 1008 female *Daphnia* with 63 individuals per  
161 treatment, in a fully factorial design (2 maternal temperatures x 2 focal temperatures x [3  
162 pathogens + uninfected controls]). For infection, individual *Daphnia* were exposed to 40,000 *P.*  
163 *ramosa* spores starting three days after birth. Pathogen exposure took place in 70-mL jars filled  
164 with 20 mL of artificial media for three days, after which all animals were transferred to 50 mL  
165 fresh media and maintained as described above.

#### 166 **Heat tolerance assays**

167 Static heat shock assays were used to measure the heat tolerance, quantified via knockdown  
168 times, of *Daphnia* from all treatments described above. Knockdown times measure the capacity of  
169 an animal to avoid physical incapacitation during thermal extremes (Hector et al., 2019; K. A.  
170 Mitchell & Hoffmann, 2010). Individual *Daphnia* were placed in 5-mL glass fly vials covered in  
171 mesh and immersed in a constantly agitated water bath filled with media and set to 37°C, which is  
172 an acute heat stress that is lethal to animals after several hours or less (Hector et al., 2019;  
173 Yampolsky et al., 2014). Starting from when they were first placed in the water bath, time until  
174 knockdown was recorded for each *Daphnia* when there was no visible movement (Hector et al.,  
175 2019; Yampolsky et al., 2014). A total of 36 *Daphnia* per treatment were chosen at random to  
176 measure heat tolerance. Three individuals per treatment could be measured per assay run, so 12  
177 runs were conducted over three consecutive days. All animals were between 19- and 21-days  
178 post-infection at the time of the assays.

#### 179 **Measuring the characteristics of individual hosts and pathogens**

180 *Daphnia* that were not used in the heat tolerance assays were kept at their respective focal  
181 acclimation temperatures until death. From birth, these animals were checked daily for deaths,  
182 and any dead animals were frozen in 500 µL of RO water for later bacterial spore counting (see  
183 below). Offspring were counted and removed twice weekly for all experimental individuals. This  
184 gave us four important metrics of individual host and pathogen performance for each temperature  
185 by pathogen treatment combination: host lifespan, host age-specific fecundity, pathogen spore  
186 loads at host death, and infection rates.

187 Bacterial spore counts were quantified using an Accuri C6 Flow Cytometer (BD Biosciences, San  
188 Jose, California). Infected animals were thawed and homogenised in 500 µL of RO water. Then, 10  
189 µL of this sample was pipetted into 190 µL of 5mM EDTA in a 96-well plate. A combination of gates  
190 based on fluorescence (via the 670 LP filter) and side scatter (cell granularity) were used to identify  
191 mature spores based on their distinct size, morphology, and fluorescence, compared to immature

192 spores, algae, or animal debris. Each sample was counted twice, and the average used to calculate  
193 total spore load per infected individual.

#### 194 **Analysis of individual-based metrics**

195 All analyses were conducted in R (v. 3.6.2; R Development Core Team, [www.R-project.com](http://www.R-project.com)). For all  
196 traits, maternal acclimation temperature (2 levels: 20°C or 25°C), focal acclimation temperature (2  
197 levels: 20°C or 25°C), pathogen treatment (4 levels: pathogen genotype C1, C14 and C20, or  
198 uninfected controls) and their interactions were fitted as fixed effects and analysed via an analysis  
199 of variance (ANOVA Type III; *car* package: (Fox & Weisberg, 2011)). For this analysis, age-specific  
200 fecundity data were summed to generate a metric of lifetime fecundity for each individual host.  
201 Due to differences in early survival, handling errors, and male individuals set up unintentionally,  
202 sample sizes for the different treatment combinations and disease traits varied between 17 and  
203 26. All exposed individuals were included in the analysis of host traits in order to capture the  
204 changes in a host's phenotype that can manifest even when infection isn't ultimately "successful"  
205 for the pathogen in producing mature transmission spores (Butterworth et al., 2024; M. D. Hall et  
206 al., 2024).

207 For heat knockdown times used a linear mixed effect model (*nlme* package; (Pinheiro & Bates,  
2008 2000) with assay run treated as a random effect. We allowed residual variance to vary  
209 independently at the level of the focal acclimation temperature to account for heteroscedasticity  
210 using the '*VarIdent*' function. Least-squared linear models were then used to analyse host lifespan,  
211 host lifetime fecundity, and pathogen spore loads, with host lifespan and fecundity both log-  
212 transformed before analysis. For spore loads, we additionally used a white-corrected analysis of  
213 variance to account for residual heteroscedasticity. Finally, for infection rates (i.e., the probability  
214 that each pathogen genotype would infect and go on to produce mature transmission spores), we  
215 used a binomial generalized linear model, with mean infection rates and standard errors extracted  
216 using *emmeans* package (see [github.com/rvlenth/emmeans](https://github.com/rvlenth/emmeans)).

#### 217 **Modelling host population growth in the absence of the pathogen**

218 To quantify the impact of thermal acclimation on the population growth potential of the host, we  
219 calculated the intrinsic rate of increase ( $r_m$ ) using lifespan and age-specific fecundity data from the  
220 unexposed (i.e., control) animals in each treatment (Fig. S1). Following the methodology of  
221 Shocket et al. (2018) we used a simplified version of the Euler-Lotka equation to calculate the  
222 intrinsic rate of increase for each individual (rather than for each whole population),



$$1 = \sum_t e^{-r_m t} l_t F_t \quad (1)$$

223 where for each single individual,  $l_t$  (i.e., the proportion of individuals in a cohort surviving to day  $t$ )  
 224 always equals 1, while the animals remain alive, and  $F_t$  is the fecundity of each individual at day  $t$ .

225 From these data, we also estimated additional demographic metrics that play an important role in  
 226 understanding the likely spread of the pathogen, which assume the density-dependence of host  
 227 birth rates (see below). We first calculated the instantaneous death rate ( $d$ ) for our control hosts  
 228 in each temperature treatment assuming time until death followed an exponential distribution,  
 229 where the likelihood of a constant death rate ( $d$ ) is calculated from our time until death (lifespan)  
 230 data under each temperature treatment (Civitello et al., 2013; Shocket, Strauss, et al., 2018) as per

$$\ell(d|t_d) = d e^{-d t_d}. \quad (2)$$

231 We then estimated the birth rate ( $b$ ) of hosts as the sum of their intrinsic rate of increase ( $r_m$ ) and  
 232 death rate ( $d$ ) for the control animals (where  $b = r_m + d$ , (Civitello et al., 2013; Shocket, Strauss, et  
 233 al., 2018)).

#### 234 **Modelling the spread and population growth of the pathogen**

235 For a pathogen, an analogous measure to the hosts intrinsic rate of increase ( $r_m$ ) is the basic  
 236 reproduction number,  $R_0$ , which informs us about a pathogens potential to spread through an  
 237 entirely susceptible population (Anderson & May, 1986). Larger values of  $R_0$  suggest the potential  
 238 for larger epidemics. For the *Daphnia-Pasteuria* disease system,  $R_0$  can be derived from a  
 239 compartmental model that tracks changes in the density of susceptible hosts, infected hosts, and  
 240 environmental pathogen spores ((S. R. Hall et al., 2009) see also (Aulsebrook et al., 2023; Civitello  
 241 et al., 2013)). From this model,  $R_0$  is calculated as

$$R_0 = \left( \frac{b - d}{bc} \right) \left( \frac{\sigma \beta}{m} \right) \quad (3)$$

242 which is conditional on the density-dependent dynamics of the host population in the absence of  
 243 disease,  $(b - d)/(bc)$ , and three epidemiological traits,  $(\sigma \beta / m)$ .  $R_0$  will increase if there are increases  
 244 in host birth rate,  $b$ , environmental transmission rate,  $\beta$ , or pathogen spore loads,  $\sigma$ .  $R_0$  decreases  
 245 if there are increases in host death rate,  $d$ , the rate of pathogen loss from the environment,  $m$ , or  
 246 the strength of density-dependence on population growth,  $c$  (Civitello et al., 2013; S. R. Hall et al.,  
 247 2009).

#### 248 **Parameterization of host and pathogen rates of increase**

For all measures of host and pathogen population growth ( $r_m$  and  $R_0$ ), we used JAGS as implemented in R (*R2jags* package: (Su & Yajima, 2009) to calculate Bayesian posterior distribution estimates for each underlying trait, following equations 1 and 3 above. Our standard JAGS settings included 75000 iterations, 30000 burn-in, thinning of 16, and 3 individual chains. We used semi-informative priors and set the Bayesian posteriors to follow the appropriate distributions for each trait following Shocket et al (2018). To calculate  $R_0$ , we first estimated the instantaneous death rate for each treatment ( $d$ ), following equation 2, and used this to estimated instantaneous birth rates ( $b$ ) for each acclimation treatment under the assumption that  $r_m$  equals the difference between instantaneous birth and death rates ( $(r_m = b - d, (McCallum, 1999))$ ). We then estimated environmental transmission rates ( $\beta$ ) using the numbers of infected and uninfected individuals from each temperature and pathogen treatment using a binomial distribution in a likelihood function to model the number of uninfected hosts in each jar, where the probability of remaining uninfected ( $P$ ) is

$$P = e^{-\beta Zt} \quad (2)$$

where  $Z$  is the density of pathogen spores and  $t$  is the length of the infection period (see (Shocket, Strauss, et al., 2018) and the supplementary material therein for details of how this likelihood function is derived). In our estimates of environmental transmission rate ( $\beta$ ), individuals were only scored as being infected if they became infected and went on to produce mature transmission spores. For the Bayesian estimate of environmental transmission rate ( $\beta$ ) and the GLM for infection probability, one treatment achieved a 100% infection rate in our experiment (pathogen C20, temperature treatment 20°C and 20°C), so to allow more reasonable point estimates and error to be calculated we adjusted this treatment to include one uninfected individual.

Finally, to calculate  $R_0$  for each pathogen and temperature treatment, we incorporated the Bayesian posterior estimates of each of the estimated parameters into our derived equation for  $R_0$  (eqn. 1). By incorporating the posterior estimates for each calculated trait in turn, we allowed the propagation of error in our estimates of each trait into our final estimates of the potential for disease spread,  $R_0$ . Two parameters that contribute to our indicator of the potential for disease spread ( $R_0$ ), the strength of density-dependence on birth rates,  $c$ , and spore degradation rate,  $m$ , were set as constants for all treatments ( $c = 0.01$  and  $m = 0.9$ , see (Civitello et al., 2013; Shocket, Vergara, et al., 2018). Whilst it is conceivable that temperature will alter the strength of density-dependence and spore degradation, and the population dynamic of the host and pathogen in nature, neither was possible to quantify in these experiments (but see (Shocket et al., 2019)).

## 280 **Results**

### 281 **Acclimation improves thermal tolerance for both uninfected and infected hosts**

282 Individuals directly exposed to 25°C showed a clear improvement in knockdown times, regardless  
283 of whether they were infected by a pathogen or not (Fig. 1 and Table 1). For example, both control  
284 and infected individuals exposed to a focal temperature of 25°C saw a two-fold increase in  
285 knockdown times compared to hosts acclimated to 20°C. However, infection by a pathogen  
286 significantly reduced the thermal tolerance of a host compared to controls, but only in the focal  
287 25°C treatments (Fig. 1, and a significant focal by pathogen treatment interaction in Table 1). Here  
288 the influence of maternal acclimation also become most apparent, with the relative difference  
289 between the control and pathogen treatment groups depending on the maternal thermal history  
290 (20°C or 25°C), leading to a three-way interaction between maternal temperature, focal  
291 temperature, and pathogen treatment in shaping overall knockdown times (Table 1).

### 292 **Warmer temperatures decrease host and pathogen individual performance**

293 Host lifespan and lifetime fecundity were considerably lower at warm focal temperatures (Fig. 2A,  
294 B). Uninfected hosts, for example, saw almost a fifty percent reduction in both traits at 25°C  
295 compared to their counterparts at 20°C. Pathogen exposure severely reduced both traits at each  
296 acclimatisation temperature, however, the greatest difference between healthy and infected  
297 hosts occurred at the standard focal temperatures (20°C), accounting for the significant  
298 interaction between pathogen exposure and focal temperature treatments (Table 1). Maternal  
299 acclimation prior to infection also subtly modified this interaction, contributing to the three-way  
300 interactions for both traits (Table 1), as the difference between healthy and infected hosts was  
301 smallest when both maternal and focal acclimation occurred at 25°C (Figure 2A, B).

302 For the pathogen, the probability of successfully infecting a host, and the resulting production of  
303 mature transmission spores, was also reduced when directly acclimated to warmer temperatures  
304 (Fig. 2C, D). However, variation in the probability of infection depended on a significant three-way  
305 interaction between both acclimation treatments and pathogen genotype (Table 1). Infection  
306 rates were significantly lower for individuals reared directly at a 25°C and it is at this temperature  
307 that differences in infection rates emerged between the pathogen genotypes. Yet the pathogen  
308 that performed best or worse depended on the maternal acclimation temperature. At a focal  
309 temperature of 25°C, for example, pathogen C20 had the lowest infection success when maternal  
310 acclimation occurred at 20°C but outperformed all other pathogen genotypes when mothers were

311 acclimated at the warmer temperature (Fig. 2C). In contrast, mature spore loads were determined  
312 by the independent effects of focal temperature and pathogen genotype (Table 1). Overall, we  
313 saw a reduction in mature transmission spores under 25°C focal acclimation, with pathogen  
314 genotype C1 generally producing more spores than C14 and then C20 (Fig. 2D).

### 315 **Thermal acclimation has opposing effects on host and pathogen population growth**

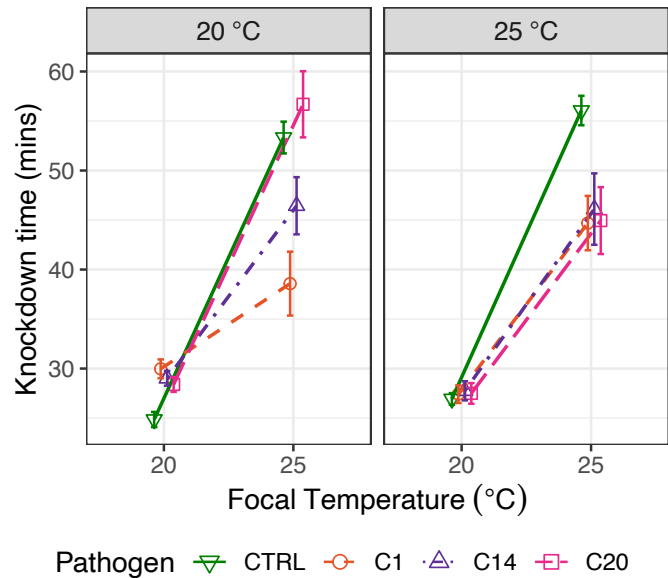
316 To predict the likely rate of population growth for the host under the different acclimation  
317 treatments, we incorporated the age-specific fecundity and lifespan of each control individual into  
318 an estimate of the intrinsic rate of increase ( $r_m$ ). In stark contrast to the negative influence of  
319 warmer temperatures on the fecundity and lifespan on individuals (Fig. 2A, B), we found that  
320 direct acclimation to temperatures promoted a statistically clear increase in  $r_m$  values (Fig. 3A,  
321 non-overlapping 95% CIs). Maternal acclimation led to a slight decrease in  $r_m$  for individuals  
322 experiencing 25°C for both the maternal and focal acclimation treatments. Overall, the potential  
323 for population growth appears to be maximised by the acceleration of early reproduction when  
324 the host directly experienced warmer temperatures (see also Fig. S1).

325 In contrast, when we estimated the pathogen's potential to spread in a susceptible population  
326 ( $R_0$ ), and thus increase in population size, we saw a considerable decrease in  $R_0$  when infecting  
327 hosts directly acclimated to 25°C (Fig. 3B, focal 25°C). Indeed, for most pathogen genotypes, the  
328 potential for disease spread was around an order of magnitude lower (with separation of 95% CIs),  
329 when the focal temperature experienced was 25°C compared to 20°C. As above, the maternal  
330 acclimation temperature experienced by hosts – before they encountered any pathogens – most  
331 notably influenced the rank order of  $R_0$  across pathogen genotypes when infections subsequently  
332 took place at 25°C (Fig. 3B).

333 This severe reduction in  $R_0$  at warmer focal temperatures appears to be driven entirely by the  
334 effects of temperature on pathogen environmental transmission rates and spore production.  
335 Environmental transmission rates (which capture the rate at which host become infected via the  
336 ingestion of free-living spores from the environment) and spore loads (which fuel the free-living  
337 spores pool) were generally higher when infections took place at 20°C, compared with 25°C (Fig.  
338 3D, F). The contributions of an increased supply of susceptible hosts, on which  $R_0$  also depends  
339 (see Eqn. 3), did not offset these within-host disadvantages. Despite the sensitivity of the host's  
340 intrinsic rate of increase to temperature ( $r_m$ , Fig. 3A), the slight increase in death rates associated  
341 with warmer focal temperatures (Fig. 3C) was not sufficiently strong enough to alter the rate at  
342 which carrying capacity might be reached (i.e.,  $(b - d)/b$ , Fig. 3E). Thus, the density-dependent

343 dynamics of the susceptible host population and the relative contribution of births and deaths in  
344 our model of disease spread,  $(b - d)/b$ , remain unaffected by thermal acclimation and unable to  
345 compensate for the poorer within-host performance of the pathogen at 25°C.

346 **Tables and figures**



347 Pathogen CTRL C1 C14 C20

348 Figure 1: The effect of thermal acclimation on heat knockdown times. Knockdown time was  
349 measured for *Daphnia* infected with one of three pathogen genotypes (C1, C14 or C20) or  
350 uninfected (CTRL). Each facet represents the maternal acclimation temperature treatment pre-  
351 infection, while the focal temperature was experienced by experimental animals from birth,  
352 including over the duration of the infection. Points represent treatment means ( $\pm$  SE).

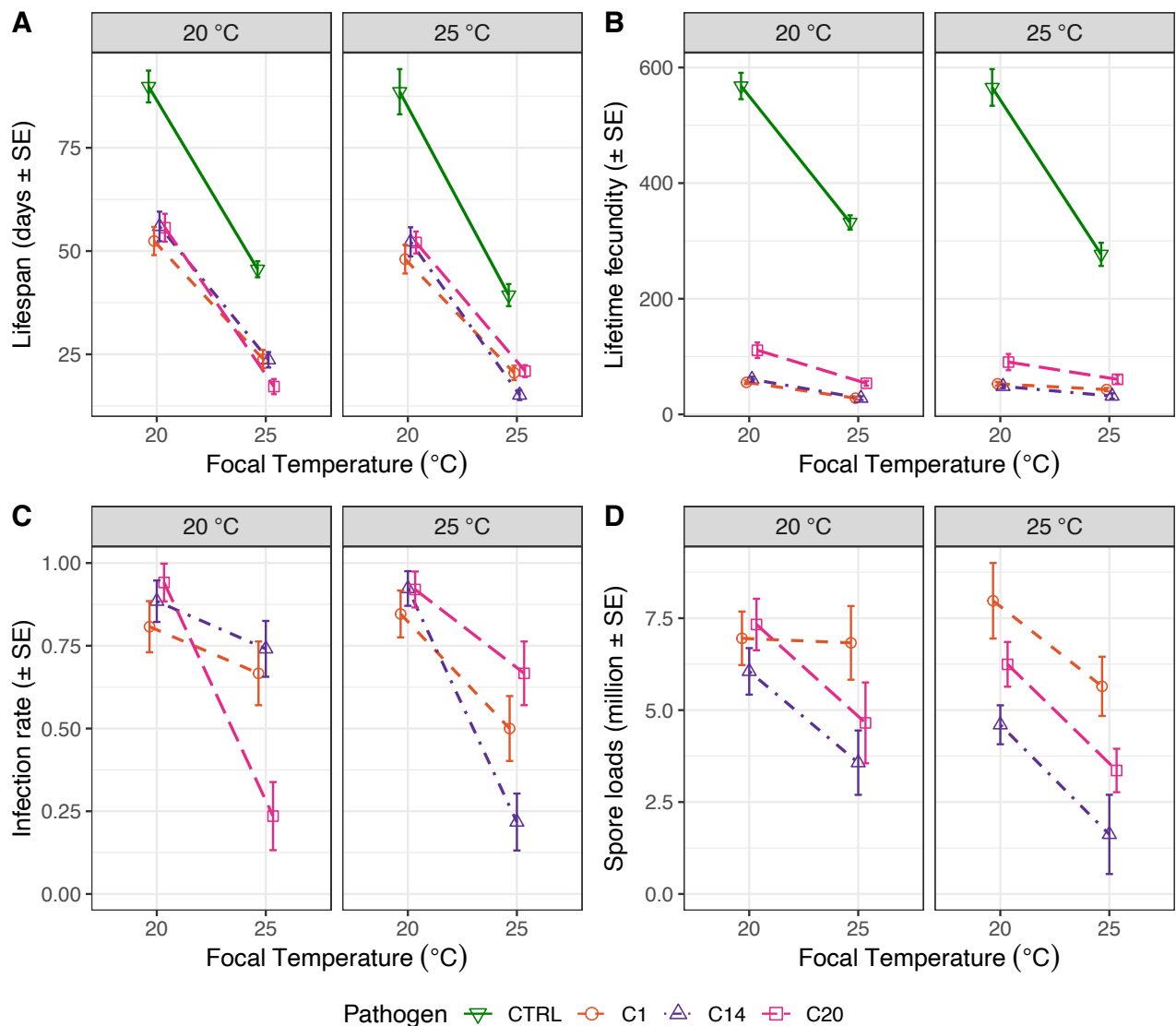
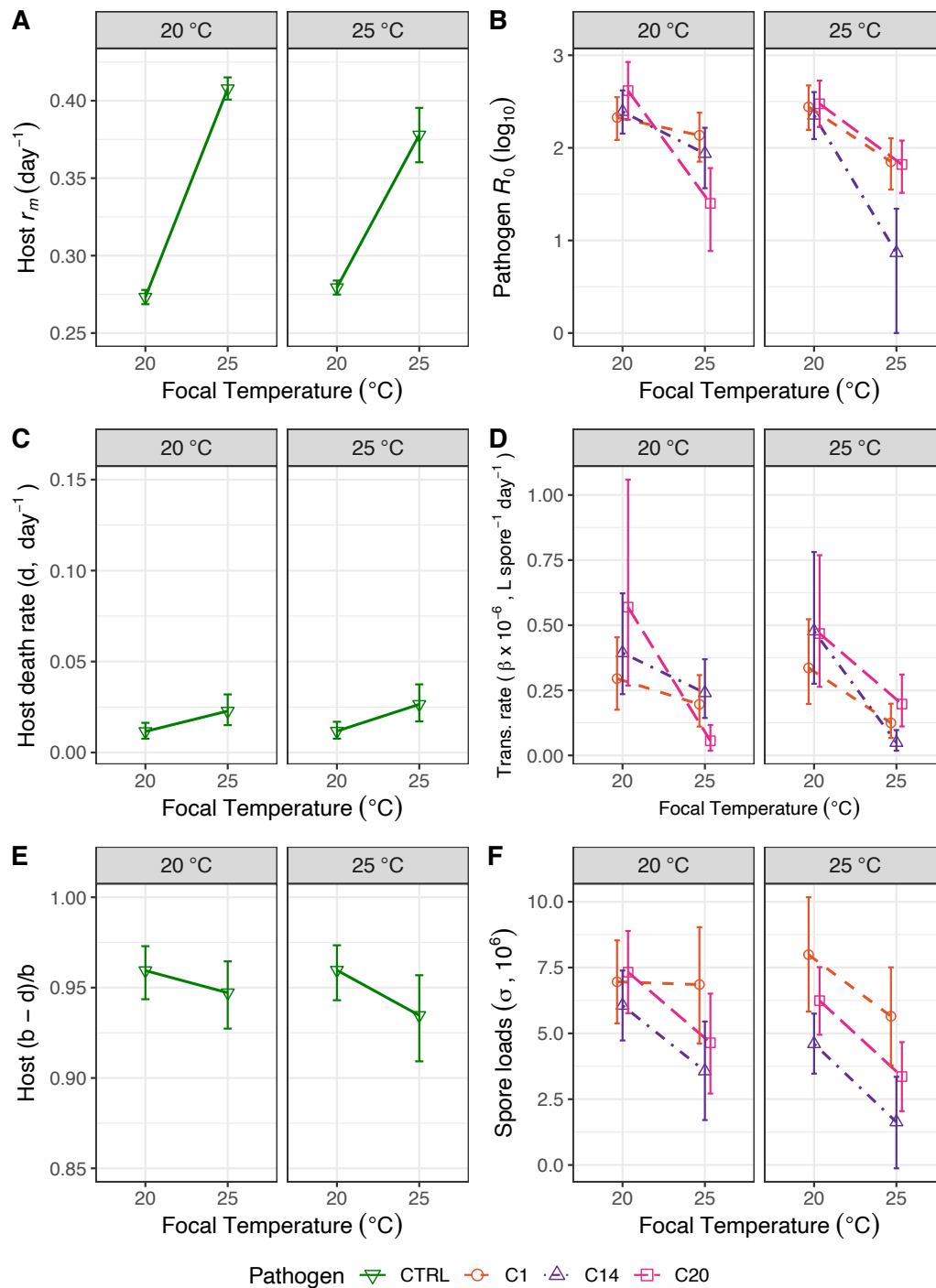


Figure 2: The effect of thermal acclimation and infection on host and pathogen life-history traits as measured for *Daphnia* exposed to one of three pathogen genotypes (C1, C14 or C20) or maintained as unexposed control animals (CTRL). Shown are the treatment means ( $\pm$  SE) for A) host lifespan, B) host fecundity over their lifetime, C) the proportion of hosts infected by the pathogen, and D) the subsequent production of mature spores at host death. Each facet represents the maternal thermal acclimation temperature. The focal temperature was experienced by experimental animals from birth, including over the duration of the infection.



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Figure 3: The effect of thermal acclimation on the predicted growth of an unexposed host ( $r_m$ ) and pathogen ( $R_0$ ) populations. Shown are the treatment means and 95% credible intervals for A) the host's intrinsic rate of increase ( $r_m$ ), B) the pathogen's basic reproductive number ( $R_0$ ), as well as the key parameters required for estimating  $R_0$ , including C) death rates of unexposed hosts, D) environmental transmission rates for hosts encountering spores in the environment E) the density-dependent control of population growth for unexposed hosts (i.e.,  $(b-d)/b$ ), , and F) spore loads at host death. For the interrelated metrics of the unexposed host population (A, C, E, in green), we maintained a relatively consistent y-axis scale in each case to facilitate relative effect size comparisons.

372 Table 1: The effects of maternal thermal acclimation (20°C or 25°C), focal acclimation (20°C or  
 373 25°C), pathogen treatment (Controls, C1, C14 or C20) and all interactions on A) host knockdown  
 374 times under 37°C static heat shock, B) host lifespan (log-transformed), C) host lifetime fecundity  
 375 (log-transformed), E) pathogen infection success probability, and F) pathogen spore loads.

Term	<i>F or <math>\chi^2</math></i>	<i>df</i>	<i>p-value</i>
<b>A) Knockdown times</b>			
Maternal acclimation	0.513	1	0.474
Focal acclimation	399.944	1	<b>&lt; 0.001</b>
Pathogen treatment	14.683	3	<b>0.002</b>
Maternal x Focal	0.005	1	0.945
Maternal x Pathogen	11.090	3	<b>0.011</b>
Focal x Pathogen	32.550	3	<b>&lt; 0.001</b>
Maternal x Focal x Pathogen	11.465	3	<b>0.009</b>
<b>B) Host lifespan</b>			
Maternal acclimation	7.028	1, 375	<b>0.008</b>
Focal acclimation	605.015	1, 375	<b>&lt; 0.001</b>
Pathogen treatment	91.240	3, 375	<b>&lt; 0.001</b>
Maternal x Focal	0.613	1, 375	0.434
Maternal x Pathogen	2.956	3, 375	<b>0.032</b>
Focal x Pathogen	4.837	3, 375	<b>0.003</b>
Maternal x Focal x Pathogen	3.160	3, 375	<b>0.025</b>
<b>C) Host lifetime fecundity</b>			
Maternal acclimation	0.263	1, 375	0.608
Focal acclimation	156.177	1, 375	<b>&lt; 0.001</b>
Pathogen treatment	729.784	3, 375	<b>&lt; 0.001</b>
Maternal x Focal	6.324	1, 375	<b>0.012</b>
Maternal x Pathogen	3.590	3, 375	<b>0.014</b>
Focal x Pathogen	1.241	3, 375	0.295
Maternal x Focal x Pathogen	4.257	3, 375	<b>0.006</b>
<b>D) Pathogen infection success</b>			
Maternal acclimation	0.135	1	0.714
Focal acclimation	49.42	1	<b>&lt; 0.001</b>
Pathogen treatment	0.292	2	0.864
Maternal x Focal	0.530	1	0.466
Maternal x Pathogen	3.005	2	0.223
Focal x Pathogen	4.573	2	0.102
Maternal x Focal x Pathogen	7.445	2	<b>0.024</b>
<b>E) Pathogen spore loads</b>			
Maternal acclimation	3.797	1, 192	0.053
Focal acclimation	19.539	1, 192	<b>&lt; 0.001</b>
Pathogen treatment	10.445	2, 192	<b>&lt; 0.001</b>
Maternal x Focal	0.913	1, 192	0.340
Maternal x Pathogen	0.846	2, 192	0.431
Focal x Pathogen	0.958	2, 192	0.386
Maternal x Focal x Pathogen	0.357	2, 192	0.700



## 376 Discussion

377 Acclimation to warmer temperatures can allow individuals to shift their thermal optima or  
378 maxima, potentially acting as a buffer against future heat stress (Rohr et al., 2018; Sgrò et al.,  
379 2016; Sinclair et al., 2016), with smaller organisms, such as parasites and pathogens, potentially  
380 benefiting more from this process owing to their smaller size (Rohr et al., 2018). We found that  
381 warm acclimation improved the heat tolerance of both healthy and pathogen-exposed hosts (Fig.  
382 1). The increase in knockdown times for warm acclimated *Daphnia* was equal, if not greater, than  
383 the variation in heat tolerance found across various geographically widespread species (Hector et  
384 al., 2020; Hoffmann et al., 2012; Lasne et al., 2018; Lush et al., 2023; Sgrò et al., 2010; Yampolsky  
385 et al., 2014). Infected hosts at warmer conditions were also more heat-resistant than both healthy  
386 and infected *Daphnia* acclimated to lower temperatures, despite pathogen exposure reducing  
387 host heat tolerance at warm temperatures (Fig. 1). Thermal acclimation, therefore, appears to  
388 better prepare both infected and uninfected animals for the pressure of extreme heat events  
389 through plastic shifts in their thermal performance.

390 For both the host and pathogen, however, the improvement in heat tolerance came with a  
391 significant cost to other measures of individual performance. Exposure to 25°C led to substantial  
392 reductions in the lifespan and lifetime fecundity of both healthy and infected individuals (Fig. 1B-  
393 C). Simultaneously, at warmer temperatures, within-host pathogen performance was also  
394 reduced, as both the probability of infection success and within-host spore loads substantially  
395 dropped at 25°C. A decline in infection rates or pathogen load at similar temperatures has been  
396 previously observed in studies of different pathogens in *Daphnia* hosts (Kirk et al., 2018; Vale &  
397 Little, 2009), as well as in other host-pathogen or vector models (Fels & Kaltz, 2006; Gehman et al.,  
398 2018; Mordecai et al., 2019; Raffel et al., 2013, 2015), suggesting that 25°C may lie above the  
399 thermal optima for infection success and spore loads in this system (c.f., (Agha et al., 2018;  
400 Shocket et al., 2019)). Our results thus highlight how thermal acclimation can have opposing  
401 effects on thermal stress resistance versus the other components of host and pathogen life-history  
402 (e.g., (Cavieres et al., 2020)).

403 This contrast between the response of heat tolerance versus other traits indicates that the  
404 damage a pathogen causes is context-dependent, and not purely predicted by temperature. The  
405 decline in host fitness due to infection, known as virulence, is normally assessed in terms of  
406 reductions in lifespan or reproduction (Cressler et al., 2016; Day, 2002). In this context, we found  
407 the greatest reductions in lifespan and fecundity relative to uninfected hosts, and thus virulence,

408 occurred at lower temperatures. Individuals exposed to a pathogen at 20°C, for example,  
409 experienced a reduction in lifespan of approximately 15 days greater than that experienced at the  
410 higher temperature (Fig 2B). In contrast, individuals acclimated at 25°C experienced up to a 15-  
411 minute reduction in their knockdown times due to infection (Fig 1). These results highlight how  
412 warming can increase one aspect of pathogen virulence via a loss of heat tolerance, but negate  
413 others related to a host's life-history. In the context of escalating heat events the most crucial  
414 component of a pathogen's virulence may well be these change made to a host's heat tolerance  
415 (Hector et al., 2023).

416 Our results so far suggest that improved heat tolerance under warming temperatures comes with  
417 costs to the individual performance of both hosts and pathogens. Yet, as we show, individual  
418 performance metrics can be misleading when population persistence instead depends on vital  
419 rates such as growth rates for a host population or the between-host spread of a pathogen (i.e.,  
420 (Mideo et al., 2008). For the host, the reduction in lifetime fecundity and lifespan (with an  
421 associated increase in intrinsic death rate, Fig. 3C) was typical of warmer temperatures increasing  
422 "pace of life" traits (see also (Adamo & Lovett, 2011; Aulsebrook et al., 2022; Debecker & Stoks,  
423 2019; Hector et al., 2021)). As a result, warming favoured earlier reproduction and led to an  
424 increase in the intrinsic rate of increase ( $r_m$ ) of the host population, rather than coming with a cost  
425 to population growth. The temperature at which  $r_m$  is highest is considered to be the optimal  
426 temperature for fitness (Amarasekare & Coutinho, 2013; Amarasekare & Savage, 2012).  
427 Acclimation to warmer temperatures thus improves host fitness both in terms of thermal  
428 tolerance and population growth, despite negatively affecting the expression of key life-history  
429 traits at the level of the individual.

430 In contrast to the host, warmer temperatures reduced the capacity of a pathogen population to  
431 expand, as the basic reproduction number,  $R_0$ , for each pathogen was around an order of  
432 magnitude lower at 25°C compared to 20°C (Fig. 3B). The decline in the capacity of the pathogen  
433 to spread between hosts was driven entirely by the reductions in environmental transmission  
434 rates (Fig. 3D) and the production of spores at host death (Fig. 3F). Lower spore loads were  
435 expected as an increase in pace-of-life cuts short the duration that a pathogen can proliferate  
436 within a host (Clerc et al., 2015; Gipson et al., 2019; M. D. Hall & Mideo, 2018). However, feeding  
437 rates were expected to increase with temperature, driving higher contact rates between hosts and  
438 pathogens, and thus higher infection rates (Shocket et al., 2019; Shocket, Strauss, et al., 2018;  
439 Shocket, Vergara, et al., 2018). Warmer temperatures may instead have afforded a host an

improved immune response ((Adamo & Lovett, 2011; Ferguson et al., 2016); but see (Raffel et al., 2006), or otherwise reduced per spore infectivity via an unknown mechanism. Changes in the supply of susceptible hosts did not offset these disadvantages. The relative combination of birth and death rate in our model meant that the control of density-dependent population growth for susceptible hosts,  $(b-d/b)$  was equivalent across all temperatures (as indicated by overlapping credible intervals in Fig. 3C).

Finally, our results suggest that when a host is faced with a pathogen, contemporary temperatures experienced during infection may swamp any carryover effects of maternal or developmental acclimation (e.g., (Sun et al., 2022)). This contrasts with terrestrial insects such as *Drosophila* where developmental temperatures drive heat tolerance and fitness (Kellermann et al., 2017; Slotsbo et al., 2016). Here, the maternal thermal environment appears to play a more subtle role in shaping which pathogen genotype performs best at any given temperature (Table 1). The largest effects of maternal acclimation in this context were revealed when the offspring of 20°C acclimated mothers subsequently experienced warm focal conditions (i.e., the 20°C maternal and 25°C focal combination). For example, the relative impact of pathogen exposure on host thermal tolerance varied across maternal acclimation treatments (Fig. 1), as did the rank order of pathogen genotypes in their infection success (Fig. 3F) and, as a result,  $R_0$  (Fig. 3B). Prior thermal environments, via maternal or developmental host effects, may thus have the potential to maintain genetic variation in pathogen populations via changes to both within- and between-host infection dynamics (Fels & Kaltz, 2006; Garbutt et al., 2014; Vale et al., 2008; Vale & Little, 2009).

In conclusion, we have shown that acclimation at warmer temperatures can buffer both hosts and pathogens alike against further heat stress, by improving the thermal tolerance of both uninfected and infected hosts – but this comes with a cost to individual trait performance. For hosts, warming caused severe reductions to overall lifespan and fecundity, but by accelerating the pace-of-life of the host, facilitated an overall increase in predicted rates of population growth and thus ultimately fitness (Amarasekare & Coutinho, 2013; Amarasekare & Savage, 2012). The outlook for a pathogen under warmer temperatures may be bleaker. Within-host pathogen success, and ultimately the potential for disease spread, was severely hampered at warmer temperatures. If true for other species, hosts may hold an advantage over pathogens in warmer and more variable environments, both in terms of their heat tolerance and the capacity to maintain stable populations (but see Shocket et al. 2019). Projections for the eco-evolutionary dynamics of host-pathogen systems could benefit from considering the joint impacts of warming and infection on multiple host and

472 pathogen traits, and how individual-level traits link to population processes (Kirk et al., 2018;  
473 Mordecai et al., 2019; Shocket et al., 2019)

474 **Acknowledgements**

475 We would like to thank L. Aulsebrook, I. Booksmythe, L. Heffernan, C. Lasne, S. Layh and J. Lush for  
476 help with laboratory work. This work was supported by funding from both Monash University and  
477 the Australian Research Council.

478 **References**

479 Adamo, S. A., & Lovett, M. M. E. (2011). Some like it hot: the effects of climate change on  
480 reproduction, immune function and disease resistance in the cricket *Gryllus texensis*. *Journal of*  
481 *Experimental Biology*, 214(12), 1997–2004. <https://doi.org/10.1242/jeb.056531>

482 Agha, R., Gross, A., Gerphagnon, M., Rohrlack, T., & Wolinska, J. (2018). Fitness and eco-  
483 physiological response of a chytrid fungal parasite infecting planktonic cyanobacteria to  
484 thermal and host genotype variation. *Parasitology*, 145(10), 1279–1286.  
485 <https://doi.org/10.1017/s0031182018000215>

486 Amarasekare, P., & Coutinho, R. M. (2013). The intrinsic growth rate as a predictor of population  
487 viability under climate warming. *Journal of Animal Ecology*, 82(6), 1240–1253.  
488 <https://doi.org/10.1111/1365-2656.12112>

489 Amarasekare, P., & Savage, V. (2012). A framework for elucidating the temperature dependence  
490 of fitness. *The American Naturalist*, 179(2), 178–191. <https://doi.org/10.1086/663677>

491 Anderson, R. M., & May, R. M. (1986). The invasion, persistence and spread of infectious diseases  
492 within animal and plant communities. *Philosophical Transactions of the Royal Society of London*  
493 *B*, 314(1167), 533–570. <https://doi.org/10.1098/rstb.1986.0072>

494 Angilletta, M. J., Steury, T. D., & Sears, M. W. (2004). Temperature, growth rate, and body size in  
495 ectotherms: fitting pieces of a life-history puzzle. *Integrative and Comparative Biology*, 44(6),  
496 498–509. <https://doi.org/10.1093/icb/44.6.498>

497 Auld, S. K., & Brand, J. (2017). Simulated climate change, epidemic size, and host evolution across  
498 host-parasite populations. *Glob. Chang. Biol.*, 23(12), 5045–5053.  
499 <https://doi.org/10.1111/gcb.13769>

500 Aulsebrook, L. C., Wong, B. B. M., & Hall, M. D. (2022). Warmer temperatures limit the effects of  
501 antidepressant pollution on life-history traits. *Proceedings of the Royal Society B: Biological*  
502 *Sciences*, 289(1968), 20212701. <https://doi.org/10.1098/rspb.2021.2701>

503 Aulsebrook, L. C., Wong, B. B. M., & Hall, M. D. (2023). Can pharmaceutical pollution alter the  
504 spread of infectious disease? A case study using fluoxetine. *Philosophical Transactions of the*  
505 *Royal Society B: Biological Sciences*, 378(1873), 20220010.  
506 <https://doi.org/10.1098/rstb.2022.0010>

507 Beaman, J. E., White, C. R., & Seebacher, F. (2016). Evolution of Plasticity: Mechanistic Link  
 508 between Development and Reversible Acclimation. *Trends Ecol. Evol.*, 31(3), 237–249.  
 509 <https://doi.org/10.1016/j.tree.2016.01.004>

510 Beck-Johnson, L. M., Nelson, W. A., Paaijmans, K. P., Read, A. F., Thomas, M. B., & Bjørnstad, O. N.  
 511 (2017). The importance of temperature fluctuations in understanding mosquito population  
 512 dynamics and malaria risk. *Royal Society Open Science*, 4(3), 160969.  
 513 <https://doi.org/10.1098/rsos.160969>

514 Blanford, J. I., Blanford, S., Crane, R. G., Mann, M. E., Paaijmans, K. P., Schreiber, K. V., & Thomas,  
 515 M. B. (2013). Implications of temperature variation for malaria parasite development across  
 516 Africa. *Scientific Reports*, 3(1), 1300. <https://doi.org/10.1038/srep01300>

517 Burton, T., Lakka, H.-K., & Einum, S. (2020). Acclimation capacity and rate change through life in  
 518 the zooplankton *Daphnia*. *Proceedings of the Royal Society B*, 287(1924), 20200189.  
 519 <https://doi.org/10.1098/rspb.2020.0189>

520 Butterworth, N. J., Heffernan, L., & Hall, M. D. (2024). Is there a sicker sex? Dose relationships  
 521 modify male–female differences in infection prevalence. *Proceedings of the Royal Society B*,  
 522 291(2014), 20232575. <https://doi.org/10.1098/rspb.2023.2575>

523 Cavieres, G., Rezende, E. L., Clavijo-Baquet, S., Alruiz, J. M., Rivera-Rebella, C., Boher, F., &  
 524 Bozinovic, F. (2020). Rapid within- and transgenerational changes in thermal tolerance and  
 525 fitness in variable thermal landscapes. *Ecology and Evolution*, 10(15), 8105–8113.  
 526 <https://doi.org/10.1002/ece3.6496>

527 Chown, S., Hoffmann, A., Kristensen, T., Angilletta, M., Stenseth, N., & Pertoldi, C. (2010). Adapting  
 528 to climate change: a perspective from evolutionary physiology. *Climate Research*, 43(1), 3–15.  
 529 <https://doi.org/10.3354/cr00879>

530 Civitello, D. J., Penczykowski, R. M., Hite, J. L., Duffy, M. A., & Hall, S. R. (2013). Potassium  
 531 stimulates fungal epidemics in *Daphnia* by increasing host and parasite reproduction. *Ecology*,  
 532 94(2), 380–388. <https://doi.org/10.1890/12-0883.1>

533 Clerc, M., Ebert, D., & Hall, M. D. (2015). Expression of parasite genetic variation changes over the  
 534 course of infection: implications of within-host dynamics for the evolution of virulence.  
 535 *Proceedings of the Royal Society B: Biological Sciences*, 282(1804), 20142820.  
 536 <https://doi.org/10.1098/rspb.2014.2820>

537 Colinet, H., Sinclair, B. J., Vernon, P., & Renault, D. (2014). Insects in fluctuating thermal  
 538 environments. *Annual Review of Entomology*, 60(1), 123–140.  
 539 <https://doi.org/10.1146/annurev-ento-010814-021017>

540 Cressler, C. E., Mcleod, D. V., ROZINS, C., Hoogen, J. V. D., & Day, T. (2016). The adaptive evolution  
 541 of virulence: a review of theoretical predictions and empirical tests. *Parasitology*, 143(7), 915–  
 542 930. <https://doi.org/10.1017/s003118201500092x>

543 Day, T. (2002). On the evolution of virulence and the relationship between various measures of  
 544 mortality. *Proceedings of the Royal Society B: Biological Sciences*, 269(1498), 1317–1323.  
 545 <https://doi.org/10.1098/rspb.2002.2021>

- 546 Debecker, S., & Stoks, R. (2019). Pace of life syndrome under warming and pollution: integrating  
547 life history, behavior, and physiology across latitudes. *Ecological Monographs*, 89(1), e01332.  
548 <https://doi.org/10.1002/ecm.1332>
- 549 Ebert, D., Duneau, D., Hall, M. D., Luijckx, P., Andras, J. P., Pasquier, L. D., & Ben-Ami, F. (2016). A  
550 population biology perspective on the stepwise infection process of the bacterial pathogen  
551 *Pasteuria ramosa* in *Daphnia*. *Advances in Parasitology*, 91, 265–310.  
552 <https://doi.org/10.1016/bs.apar.2015.10.001>
- 553 Ebert, D., Zschokke-Rohringer, C. D., & Carius, H. J. (1998). Within–and between–population  
554 variation for resistance of *Daphnia magna* to the bacterial endoparasite *Pasteuria ramosa*.  
555 *Proceedings of the Royal Society B: Biological Sciences*, 265(1410), 2127–2134.  
556 <https://doi.org/10.1098/rspb.1998.0549>
- 557 Fels, D., & Kaltz, O. (2006). Temperature-dependent transmission and latency of *Holosporea*  
558 *undulata*, a micronucleus-specific parasite of the ciliate *Paramecium caudatum*. *Proceedings of*  
559 *the Royal Society B: Biological Sciences*, 273(1589), 1031–1038.  
560 <https://doi.org/10.1098/rspb.2005.3404>
- 561 Ferguson, L. V., Heinrichs, D. E., & Sinclair, B. J. (2016). Paradoxical acclimation responses in the  
562 thermal performance of insect immunity. *Oecologia*, 181(1), 77–85.  
563 <https://doi.org/10.1007/s00442-015-3529-6>
- 564 Ferguson, L. V., & Sinclair, B. J. (2019). Thermal variability and plasticity drive the outcome of a  
565 host-pathogen interaction. *The American Naturalist*, 195(4), 603–615.  
566 <https://doi.org/10.1086/707545>
- 567 Fox, J., & Weisberg, S. (2011). *An R Companion to Applied Regression*. SAGE.  
568 [http://books.google.com.au/books?id=YH6NotdvzF0C&printsec=frontcover&dq=An+R+Compa](http://books.google.com.au/books?id=YH6NotdvzF0C&printsec=frontcover&dq=An+R+Companion+to+Applied+Regression+2011&hl=&cd=1&source=gbs_api)  
569 [nion+to+Applied+Regression+2011&hl=&cd=1&source=gbs\\_api](http://books.google.com.au/books?id=YH6NotdvzF0C&printsec=frontcover&dq=An+R+Companion+to+Applied+Regression+2011&hl=&cd=1&source=gbs_api)
- 570 Garbutt, J. S., Scholefield, J. A., Vale, P. F., & Little, T. J. (2014). Elevated maternal temperature  
571 enhances offspring disease resistance in *Daphnia magna*. *Functional Ecology*, 28(2), 424–431.  
572 <https://doi.org/10.1111/1365-2435.12197>
- 573 Gehman, A.-L. M., Hall, R. J., & Byers, J. E. (2018). Host and parasite thermal ecology jointly  
574 determine the effect of climate warming on epidemic dynamics. *Proceedings of the National*  
575 *Academy of Sciences*, 115(4), 744–749. <https://doi.org/10.1073/pnas.1705067115>
- 576 Gipson, S. A. Y., Jimenez, L., & Hall, M. D. (2019). Host sexual dimorphism affects the outcome of  
577 within-host pathogen competition. *Evolution*, 73(7), 1443–1455.  
578 <https://doi.org/10.1111/evo.13760>
- 579 Greenspan, S. E., Bower, D. S., Roznik, E. A., Pike, D. A., Marantelli, G., Alford, R. A., Schwarzkopf,  
580 L., & Scheffers, B. R. (2017). Infection increases vulnerability to climate change via effects on  
581 host thermal tolerance. *Scientific Reports*, 7(1), 9349. [https://doi.org/10.1038/s41598-017-](https://doi.org/10.1038/s41598-017-09950-3)  
582 [09950-3](https://doi.org/10.1038/s41598-017-09950-3)

583 Hall, M. D., & Mideo, N. (2018). Linking sex differences to the evolution of infectious disease life-  
584 histories. *Philosophical Transactions of the Royal Society B: Biological Sciences*, 373(1757),  
585 20170431. <https://doi.org/10.1098/rstb.2017.0431>

586 Hall, M. D., Phillips, B. L., White, C. R., & Marshall, D. J. (2024). The hidden costs of resistance:  
587 Contrasting the energetics of successfully and unsuccessfully fighting infection. *Functional*  
588 *Ecology*. <https://doi.org/10.1111/1365-2435.14523>

589 Hall, M. D., Routtu, J., & Ebert, D. (2019). Dissecting the genetic architecture of a stepwise  
590 infection process. *Molecular Ecology*, 28(17), 3942–3957. <https://doi.org/10.1111/mec.15166>

591 Hall, S. R., Knight, C. J., Becker, C. R., Duffy, M. A., Tessier, A. J., & Cáceres, C. E. (2009). Quality  
592 matters: resource quality for hosts and the timing of epidemics. *Ecology Letters*, 12(2), 118–  
593 128. <https://doi.org/10.1111/j.1461-0248.2008.01264.x>

594 Hector, T. E., Gehman, A.-L. M., & King, K. C. (2023). Infection burdens and virulence under heat  
595 stress: ecological and evolutionary considerations. *Philosophical Transactions of the Royal*  
596 *Society B*, 378(1873), 20220018. <https://doi.org/10.1098/rstb.2022.0018>

597 Hector, T. E., Sgrò, C. M., & Hall, M. D. (2019). Pathogen exposure disrupts an organism's ability to  
598 cope with thermal stress. *Global Change Biology*, 25(11), 3893–3905.  
599 <https://doi.org/10.1111/gcb.14713>

600 Hector, T. E., Sgrò, C. M., & Hall, M. D. (2020). The influence of immune activation on thermal  
601 tolerance along a latitudinal cline. *Journal of Evolutionary Biology*, 33(9), 1224–1234.  
602 <https://doi.org/10.1111/jeb.13663>

603 Hector, T. E., Sgrò, C. M., & Hall, M. D. (2021). Temperature and pathogen exposure act  
604 independently to drive host phenotypic trajectories. *Biology Letters*, 17(6), 20210072.  
605 <https://doi.org/10.1098/rsbl.2021.0072>

606 Hoffmann, A. A., Chown, S. L., & Clusella-Trullas, S. (2012). Upper thermal limits in terrestrial  
607 ectotherms: how constrained are they? *Functional Ecology*, 27(4), 934–949.  
608 <https://doi.org/10.1111/j.1365-2435.2012.02036.x>

609 Kellermann, V., Heerwaarden, B. van, & Sgrò, C. M. (2017). How important is thermal history?  
610 Evidence for lasting effects of developmental temperature on upper thermal limits in  
611 *Drosophila melanogaster*. *Proceedings of the Royal Society B: Biological Sciences*, 284(1855),  
612 20170447. <https://doi.org/10.1098/rspb.2017.0447>

613 Kirk, D., Jones, N., Peacock, S., Phillips, J., Molnár, P. K., Krkošek, M., & Luijckx, P. (2018). Empirical  
614 evidence that metabolic theory describes the temperature dependency of within-host parasite  
615 dynamics. *PLOS Biology*, 16(2), e2004608. <https://doi.org/10.1371/journal.pbio.2004608>

616 Kunze, C., Luijckx, P., Jackson, A. L., & Donohue, I. (2022). Alternate patterns of temperature  
617 variation bring about very different disease outcomes at different mean temperatures. *ELife*,  
618 11, e72861. <https://doi.org/10.7554/elife.72861>



- 619 Laidlaw, T., Hector, T. E., Sgrò, C. M., & Hall, M. D. (2020). Pathogen exposure reduces sexual  
620 dimorphism in a host's upper thermal limits. *Ecology and Evolution*, 10(23), 12851–12859.  
621 <https://doi.org/10.1002/ece3.6828>
- 622 Lasne, C., Hangartner, S. B., Connallon, T., & Sgrò, C. M. (2018). Cross-sex genetic correlations and  
623 the evolution of sex-specific local adaptation: Insights from classical trait clines in *Drosophila*  
624 *melanogaster*. *Evolution*, 72(6), 1317–1327. <https://doi.org/10.1111/evo.13494>
- 625 Lush, J., Sgrò, C. M., & Hall, M. D. (2023). Anticipating change: the impact of simulated seasonal  
626 heterogeneity on heat tolerances along a latitudinal cline. *BioRxiv*.
- 627 McCallum, H. (1999). *Population Parameters: Estimation for Ecological Models*.  
628 <https://doi.org/10.1002/9780470757468>
- 629 Mideo, N., Alizon, S., & Day, T. (2008). Linking within- and between-host dynamics in the  
630 evolutionary epidemiology of infectious diseases. *Trends in Ecology & Evolution*, 23(9), 511–  
631 517. <https://doi.org/10.1016/j.tree.2008.05.009>
- 632 Mitchell, K. A., & Hoffmann, A. A. (2010). Thermal ramping rate influences evolutionary potential  
633 and species differences for upper thermal limits in *Drosophila*. *Functional Ecology*, 24(3), 694–  
634 700. <https://doi.org/10.1111/j.1365-2435.2009.01666.x>
- 635 Mitchell, S. E., Rogers, E. S., Little, T. J., & Read, A. F. (2005). Host-parasite and genotype-by-  
636 environment interactions: temperature modifies potential for selection by a sterilizing  
637 pathogen. *Evolution*, 59(1), 70–80. <https://doi.org/10.1554/04-526>
- 638 Mordecai, E. A., Caldwell, J. M., Grossman, M. K., Lippi, C. A., Johnson, L. R., Neira, M., Rohr, J. R.,  
639 Ryan, S. J., Savage, V., Shocket, M. S., Sippy, R., Ibarra, A. M. S., Thomas, M. B., & Villena, O.  
640 (2019). Thermal biology of mosquito-borne disease. *Ecology Letters*, 22(10), 1690–1708.  
641 <https://doi.org/10.1111/ele.13335>
- 642 Paaijmans, K. P., Read, A. F., & Thomas, M. B. (2009). Understanding the link between malaria risk  
643 and climate. *Proceedings of the National Academy of Sciences of the United States of America*,  
644 106(33), 13844–13849. <https://doi.org/10.1073/pnas.0903423106>
- 645 Paull, S. H., Raffel, T. R., LaFonte, B. E., & Johnson, P. T. J. (2015). How temperature shifts affect  
646 parasite production: testing the roles of thermal stress and acclimation. *Functional Ecology*,  
647 29(7), 941–950. <https://doi.org/10.1111/1365-2435.12401>
- 648 Penczykowski, R. M., Laine, A., & Koskella, B. (2016). Understanding the ecology and evolution of  
649 host–parasite interactions across scales. *Evolutionary Applications*, 9(1), 37–52.  
650 <https://doi.org/10.1111/eva.12294>
- 651 Pinheiro, J. C., & Bates, D. M. (2000). *Mixed-Effects Models in S and S-PLUS*. Springer, New York,  
652 NY. <https://doi.org/10.1007/b98882>
- 653 Porras, M. F., Agudelo-Cantero, G. A., Santiago-Martínez, M. G., Navas, C. A., Loeschcke, V.,  
654 Sørensen, J. G., & Rajotte, E. G. (2021). Fungal infections lead to shifts in thermal tolerance and  
655 voluntary exposure to extreme temperatures in both prey and predator insects. *Scientific*  
656 *Reports*, 11(1), 21710. <https://doi.org/10.1038/s41598-021-00248-z>



657 Raffel, T. R., Halstead, N. T., McMahon, T. A., Davis, A. K., & Rohr, J. R. (2015). Temperature  
658 variability and moisture synergistically interact to exacerbate an epizootic disease. *Proceedings*  
659 *of the Royal Society B: Biological Sciences*, 282(1801), 20142039.  
660 <https://doi.org/10.1098/rspb.2014.2039>

661 Raffel, T. R., Rohr, J. R., Kiesecker, J. M., & Hudson, P. J. (2006). Negative effects of changing  
662 temperature on amphibian immunity under field conditions. *Functional Ecology*, 20(5), 819–  
663 828. <https://doi.org/10.1111/j.1365-2435.2006.01159.x>

664 Raffel, T. R., Romansic, J. M., Halstead, N. T., McMahon, T. A., Venesky, M. D., & Rohr, J. R. (2013).  
665 Disease and thermal acclimation in a more variable and unpredictable climate. *Nature Climate*  
666 *Change*, 3(2), 146–151. <https://doi.org/10.1038/nclimate1659>

667 Rohr, J. R., Civitello, D. J., Cohen, J. M., Roznik, E. A., Sinervo, B., & Dell, A. I. (2018). The complex  
668 drivers of thermal acclimation and breadth in ectotherms. *Ecology Letters*, 21(9), 1425–1439.  
669 <https://doi.org/10.1111/ele.13107>

670 Rohr, J. R., & Raffel, T. R. (2010). Linking global climate and temperature variability to widespread  
671 amphibian declines putatively caused by disease. *Proceedings of the National Academy of*  
672 *Sciences*, 107(18), 8269–8274. <https://doi.org/10.1073/pnas.0912883107>

673 Sgrò, C. M., Overgaard, J., Kristensen, T. N., Mitchell, K. A., Cockerell, F. E., & Hoffmann, A. A.  
674 (2010). A comprehensive assessment of geographic variation in heat tolerance and hardening  
675 capacity in populations of *Drosophila melanogaster* from eastern Australia. *Journal of*  
676 *Evolutionary Biology*, 23(11), 2484–2493. <https://doi.org/10.1111/j.1420-9101.2010.02110.x>

677 Sgrò, C. M., Terblanche, J. S., & Hoffmann, A. A. (2016). What can plasticity contribute to insect  
678 responses to climate change? *Annual Review of Entomology*, 61(1), 1–19.  
679 <https://doi.org/10.1146/annurev-ento-010715-023859>

680 Shocket, M. S., Magnante, A., Duffy, M. A., Cáceres, C. E., & Hall, S. R. (2019). Can hot  
681 temperatures limit disease transmission? A test of mechanisms in a zooplankton–fungus  
682 system. *Functional Ecology*, 33(10), 2017–2029. <https://doi.org/10.1111/1365-2435.13403>

683 Shocket, M. S., Ryan, S. J., & Mordecai, E. A. (2018). Temperature explains broad patterns of Ross  
684 River virus transmission. *Elife*, 7, e37762. <https://doi.org/10.7554/elife.37762>

685 Shocket, M. S., Strauss, A. T., Hite, J. L., Šljivar, M., Civitello, D. J., Duffy, M. A., Cáceres, C. E., &  
686 Hall, S. R. (2018). Temperature drives epidemics in a zooplankton–fungus disease system: a  
687 trait-driven approach points to transmission via host foraging. *American Naturalist*, 191(4),  
688 435–451. <https://doi.org/10.1086/696096>

689 Shocket, M. S., Vergara, D., Sickbert, A. J., Walsman, J. M., Strauss, A. T., Hite, J. L., Duffy, M. A.,  
690 Cáceres, C. E., & Hall, S. R. (2018). Parasite rearing and infection temperatures jointly influence  
691 disease transmission and shape seasonality of epidemics. *Ecology*, 99(9), 1975–1987.  
692 <https://doi.org/10.1002/ecy.2430>

693 Sinclair, B. J., Marshall, K. E., Sewell, M. A., Levesque, D. L., Willett, C. S., Slotsbo, S., Dong, Y.,  
694 Harley, C. D. G., Marshall, D. J., Helmuth, B. S., & Huey, R. B. (2016). Can we predict ectotherm

695 responses to climate change using thermal performance curves and body temperatures?  
696 *Ecology Letters*, 19(11), 1372–1385. <https://doi.org/10.1111/ele.12686>

697 Slotsbo, S., Schou, M. F., Kristensen, T. N., Loeschcke, V., & Sørensen, J. G. (2016). Reversibility of  
698 developmental heat and cold plasticity is asymmetric and has long-lasting consequences for  
699 adult thermal tolerance. *Journal of Experimental Biology*, 219(17), 2726–2732.  
700 <https://doi.org/10.1242/jeb.143750>

701 Somero, G. N. (2010). The physiology of climate change: how potentials for acclimatization and  
702 genetic adaptation will determine ‘winners’ and ‘losers.’ *Journal of Experimental Biology*,  
703 213(6), 912–920. <https://doi.org/10.1242/jeb.037473>

704 Su, Y.-S., & Yajima, M. (2009). *R2jags: using R to Run ‘JAGS’. R package version 0.5-7. See*  
705 <https://CRAN.Rproject.org/package=R2jags> (R package version 0.5-7.) [Http://CRAN. R-project.  
706 org/package= R2jags].

707 Sun, S.-J., Dziuba, M. K., McIntire, K. M., Jaye, R. N., & Duffy, M. A. (2022). Transgenerational  
708 plasticity alters parasite fitness in changing environments. *Parasitology*, 149(11), 1515–1520.  
709 <https://doi.org/10.1017/s0031182022001056>

710 Thomas, M. B., & Blanford, S. (2003). Thermal biology in insect-parasite interactions. *Trends in*  
711 *Ecology & Evolution*, 18(7), 344–350. [https://doi.org/10.1016/s0169-5347\(03\)00069-7](https://doi.org/10.1016/s0169-5347(03)00069-7)

712 Vale, P. F., & Little, T. J. (2009). Measuring parasite fitness under genetic and thermal variation.  
713 *Heredity*, 103(2), 102–109. <https://doi.org/10.1038/hdy.2009.54>

714 Vale, P. F., Stjernman, M., & Little, T. J. (2008). Temperature-dependent costs of parasitism and  
715 maintenance of polymorphism under genotype-by-environment interactions. *Journal of*  
716 *Evolutionary Biology*, 21(5), 1418–1427. <https://doi.org/10.1111/j.1420-9101.2008.01555.x>

717 Vázquez, D. P., Gianoli, E., Morris, W. F., & Bozinovic, F. (2015). Ecological and evolutionary  
718 impacts of changing climatic variability: Impacts of changing climatic variability. *Biological*  
719 *Reviews*, 92(1), 22–42. <https://doi.org/10.1111/brv.12216>

720 Ware-Gilmore, F., Sgrò, C. M., Xi, Z., Dutra, H. L. C., Jones, M. J., Shea, K., Hall, M. D., Thomas, M.  
721 B., & McGraw, E. A. (2021). Microbes increase thermal sensitivity in the mosquito *Aedes*  
722 *aegypti*, with the potential to change disease distributions. *PLOS Neglected Tropical Diseases*,  
723 15(7), e0009548. <https://doi.org/10.1371/journal.pntd.0009548>

724 Yampolsky, L. Y., Schaer, T. M. M., & Ebert, D. (2014). Adaptive phenotypic plasticity and local  
725 adaptation for temperature tolerance in freshwater zooplankton. *Proceedings of the Royal*  
726 *Society B: Biological Sciences*, 281(1776), 20132744. <https://doi.org/10.1098/rspb.2013.2744>

727