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Mosquito thermal tolerance is remarkably constrained across a large climatic range

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21 Abstract

22

23 How mosquitoes may respond to rapid climate warming remains unknown for most
24 species, but will have major consequences for their future distributions, with cascading impacts
25 on human well-being, biodiversity, and ecosystem function. We investigated the adaptive
26 potential of a wide-ranging mosquito species, *Aedes sierrensis*, across a large climatic gradient
27 by conducting a common garden experiment measuring the thermal limits of mosquito life
28 history traits. Although field-collected populations originated from vastly different thermal
29 environments that spanned over 1,200 km, we found remarkably limited variation in upper
30 thermal tolerance between populations, with the upper thermal limits of fitness varying by $<1^{\circ}\text{C}$
31 across the species range. For one life history trait—pupal development rate—we did detect
32 significant variation in upper thermal limits between populations, and this variation was strongly
33 correlated with source temperatures, providing evidence of local thermal adaptation for pupal
34 development. However, we found environmental temperatures already regularly exceed our
35 highest estimated upper thermal limits throughout most of the species range, suggesting limited
36 potential for mosquito thermal tolerance to evolve on pace with warming. Strategies for avoiding
37 high temperatures such as diapause, phenological shifts, and behavioral thermoregulation are
38 likely important for mosquito persistence.

39

40 Introduction

41

42 How mosquitoes respond in the face of rapid anthropogenic climate warming is a key open
43 question of ecological and public health concern. As temperature impacts nearly all aspects of
44 mosquito life cycles, climate warming may cause large shifts in their distributions and
45 dynamics^{1,2}. In particular, current predictions suggest that mosquito distributions may shift
46 higher in latitude and elevation, expanding into temperate regions as they become newly
47 suitable, and contracting in some tropical regions as they become too warm³⁻⁶. However, these
48 predictions have not typically incorporated the potential for mosquito adaptive responses, and
49 thus may overestimate declines at current warm edges.

50 Temperature sets fundamental limits on mosquito distributions as mosquito survival and
51 reproduction are inhibited beyond critical thermal limits. As temperatures exceed those limits
52 under warming, mosquito populations could persist through a variety of mechanisms including
53 range shifts to track suitable temperatures, shifts in daily and/or seasonal activity patterns to
54 avoid high temperatures, behavioral thermoregulation (*i.e.*, actively seeking out cooler
55 microhabitats), and increased heat tolerance through evolutionary adaptation⁷. Of these
56 responses, evolutionary adaptation may be particularly important for enabling long-term
57 persistence, but the potential for mosquito thermal adaptation remains poorly understood, owing
58 to several empirical knowledge gaps⁸⁻¹⁰.

59 A key component of whether a given mosquito species can evolutionarily adapt to
60 warming is the presence of standing variation in upper thermal tolerance within a species¹⁰.

61 Decades of research on mosquito thermal biology have demonstrated variation in thermal
62 performance between species (*e.g.*^{1,11,12}). Further, several studies have identified within-species
63 variation in response to other aspects of climate, such as cold tolerance in *Aedes albopictus*^{13,14}
64 and aridity tolerance in *Anopheles gambiae*^{15,16}. Only a few studies have investigated within-
65 species variation in upper thermal tolerance, and have generally found some evidence of standing
66 variation (*i.e.*, differing rates of survival, reproduction, or development among populations at
67 high temperatures), but little evidence of local *thermal adaptation* (*i.e.*, higher heat tolerance
68 observed in populations from warmer environments than those from cooler environments)¹⁷⁻²¹.
69 However, these studies typically investigated relatively few mosquito populations from a limited
70 portion of the species range, owing to logistical challenges of collecting, rearing, and
71 experimenting on many wide-ranging populations. Further, mosquito thermal tolerance was
72 typically measured on select life history traits or metabolic rates, potentially obscuring patterns
73 of thermal adaptation evident across the full life cycle^{9,22,23}. Thus, the extent of variation in upper
74 thermal tolerance among populations within a species and the evidence for thermal adaptation is
75 still unknown.

76 We set out to rigorously investigate the evidence for mosquito thermal adaptation by
77 using *Aedes sierrensis*, the western tree hole mosquito, as a novel model system. *Ae. sierrensis*
78 makes an ideal model species for this investigation because it is commonly occurring across its
79 distribution (ranging from Southern California to British Columbia and coastal to montane
80 environments^{24,25}), which covers a large range of thermal environments, presenting varying
81 selection pressures and opportunities for local thermal adaptation. This species has a seasonal life
82 cycle driven by temperature, precipitation, and day length cues, and which occurs in discrete,
83 easy-to-sample habitat (water-filled tree holes)²⁵, facilitating field collection of individuals at the
84 same life stage across the species range. Further, although *Ae. sierrensis* is not a known vector of
85 human pathogens, it is congeneric to major human disease vectors (*i.e.*, *Ae. aegypti*, *Ae.*
86 *albopictus*) and is itself a vector of dog heartworm, making results potentially informative for
87 understanding warming responses in these vector species. Leveraging this model system, we set
88 out to answer the following specific research questions: (i) How much does thermal tolerance
89 vary between populations across the species range? (ii) Is variation in thermal tolerance, if
90 observed, correlated with the source thermal environment? (*i.e.*, is there evidence of local
91 thermal adaptation?) (iii) At present, how often do environmental temperatures exceed mosquito
92 populations' upper thermal limits?

93 To answer these questions, we conducted a common garden experiment using ten *Ae.*
94 *sierrensis* populations spanning nearly the entire species range (1,200 km; Figure 1). The thermal
95 environments of collected populations varied widely, with annual mean temperatures varying by
96 >7°C, and average daily maxima in the spring and summer varying by >5°C. We reared these
97 field-collected populations in the lab for one generation at common temperatures, then separated
98 F1 individuals into one of six temperature treatments ranging from 5-32°C. We tracked
99 individuals daily to approximate individual-level fitness, as well as its component life history
100 traits—larval and pupal survival and development rates, adult lifespan, and wing length (a proxy

101 for fecundity). We then fit thermal performance curves to these experimental data to estimate
102 upper and lower thermal limits, thermal optima, and breadth, and maximum performance for
103 each population and trait. In our investigation of variation in mosquito thermal tolerance, we
104 compared variation in these estimated upper thermal limits for each trait and population. We note
105 that prior studies of mosquito thermal tolerance have used a variety of methods to measure
106 thermal tolerance including static and dynamic heat tolerance assays (*e.g.*, ‘thermal
107 knockdowns’)^{12,26}, reciprocal transplants¹³, and comparisons of niche-based distribution
108 models²⁷. These methods may each capture a slightly different component of thermal tolerance
109 (*e.g.*, capacity for heat shock responses, combined genetic and plastic responses), thus our metric
110 of thermal tolerance may not be comparable across all approaches. We focused on upper thermal
111 limits from trait thermal performance curves as they capture high temperature constraints across
112 the life span.

113 Despite originating from a wide range of thermal environments, populations differed very little
114 in their thermal limits for fitness, and for nearly all other life history traits. For pupal
115 development rate, we did find significant variation in upper thermal limits between populations,
116 with five times greater variation in upper thermal limits than previously found in ectotherm
117 species across this same range. Further, this variation corresponded with populations’ source
118 thermal environments, providing evidence of local thermal adaptation. However, environmental
119 temperatures across most of the species range already regularly exceed populations’ estimated
120 upper thermal limits, suggesting thermal adaptation alone may play a limited role in enabling
121 persistence under warming. Seasonal life history strategies and behavioral thermoregulation are
122 likely important strategies for mosquitoes coping with ongoing climate warming.

123

124

125 **Methods**

126

127 *Field collection*

128 *Ae. sierrensis* typically completes one life cycle per year, with adults laying eggs in naturally
129 occurring tree holes. Eggs hatch when the tree holes fill with water beginning in the late fall and
130 advance through four larval instars and one pupal life stage throughout the winter before eclosing
131 as adults in the spring and summer²⁴. Most North American *Ae. sierrensis* populations (*e.g.*,
132 those from 26-46°N), including all of our collected populations, undergo diapause between the
133 fourth larval instar and pupal life stage, and all populations undergo embryonic diapause²⁸. We
134 collected larval *Ae. sierrensis* from 346 tree holes spanning over 1,200 km across the Western
135 U.S. between October 2021 and April 2022 (Figure 1, Supplemental Table S1 for collection
136 metadata). We collected *Ae. sierrensis* and tree hole water in plastic cups and maintained these at
137 cold temperatures (< 10°C) during transportation to the lab, then at 4°C until processing. We
138 visually inspected individuals from each sampled tree hole for the presence of *Lambornella*
139 *clarki*—a ciliate parasite that can infect larval *Ae. sierrensis*. Only larvae from tree holes without
140 the parasite were used in this experiment. Further, to maintain sufficient genetic variation and

141 avoid excessive inbreeding, we reared only larvae from tree holes with at least 30 collected
142 individuals.

143

144 *Lab rearing*

145 After processing, we maintained select populations (*i.e.*, those from tree holes with ≥ 30
146 individuals and no *L. clarki*) under shared lab conditions of 21.5°C, and a 13 h: 11 h light:dark
147 cycle. We periodically fed larvae a finely-ground mix of Tetramin fish flakes (48% by weight),
148 guinea pig chow (48%), and liver powder (2%). Once reaching the adult stage, we housed
149 populations in 8 x 8 x 8 cm aluminum collapsible cages (BioQuip, Rancho Dominguez, CA,
150 USA) with continuous access to a 10% sugar solution. We offered each population a blood meal
151 of defibrinated sheep's blood approximately once per week and placed an oviposition cup,
152 consisting of a paper cup lined with water-soaked coffee filter paper, inside each cage within
153 four days of the first blood-feeding. We collected eggs and held these at room temperature for
154 two weeks, then in the refrigerator at 4°C and near 24 h darkness to mimic winter conditions and
155 promote hatching (potentially because these cold, dark conditions cause eggs to enter and exit
156 diapause, as would occur in the field; pers. comm. Bret Barner, Solano County vector control),
157 which occurred 1-3 months later.

158 To ensure sufficient sample sizes for each treatment of the experiment, we only used
159 populations that produced >300 eggs in total. This resulted in 10 populations for use in the
160 experiment (Figure 1), wherein 'population' refers to a group of individuals originating from the
161 same tree hole. These collections are highly likely to represent distinct populations, as the
162 minimum distance between any pair of populations used in the experiment was 3.4 km, and *Ae.*
163 *sierrensis* adults are weak fliers and typically do not disperse far from their larval tree hole²⁹. We
164 note that a more precise definition of a population would incorporate specific dispersal
165 capabilities and/or genetic structuring, but this has not yet been investigated for *Ae. sierrensis*.

166 To hatch eggs, we prepared a separate tray for each population, which consisted of 500
167 mL Arrowhead distilled water, 300 mL autoclaved tree hole water (combined from all sampled
168 tree holes), and ¾ tsp Brewers' yeast. We submerged egg papers from each population in trays
169 between July 4 - 6, 2022, 24 h after the respective hatching tray was prepared.

170 We note that by using F1 individuals in our experiment, we have not eliminated
171 maternal/cross-generational effects, which may impact thermal tolerance³⁰. That is, while we
172 sought to minimize direct environmental effects on thermal tolerance (*i.e.*, 'phenotypic
173 plasticity') and capture genetically-based differences, environmental effects from prior
174 generations could still impact F1 thermal tolerance.

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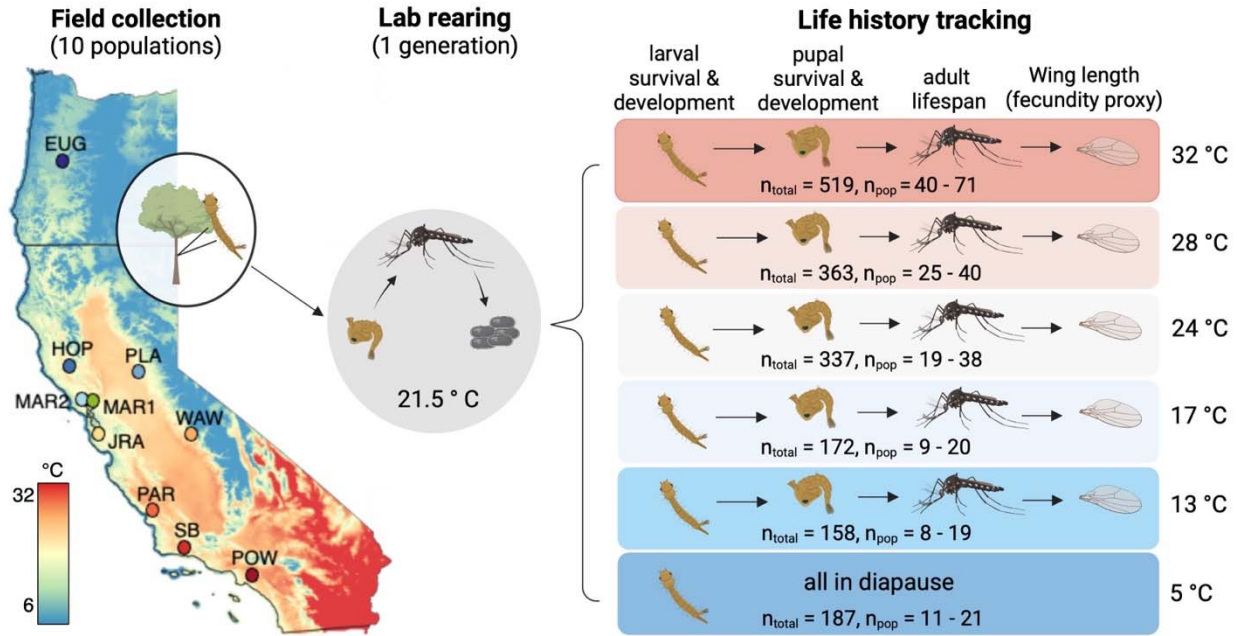
176 *Experimental design*

177 The experiment consisted of tracking life histories for individual *Ae. sierrensis* from one of ten
178 populations, held at one of six temperature treatments (Figure 1; see Supplemental Table S2 for
179 sample sizes). The temperature treatments—5, 13, 17, 24, 28, and 32°C—were chosen based on the
180 range of temperatures realistically experienced by *Ae. sierrensis* in the field and based on

181 survival rates assessed during pilot experiments conducted in the lab (Figure 1, Supplemental
182 Figure S1). These constant temperatures were maintained using Fisher Scientific Isotemp
183 incubators (for the 13, 24, 28, and 32 °C treatments) and climate-controlled rooms (for the 5 and
184 17 °C treatments). Although fluctuating temperatures could have more closely mimicked natural
185 conditions, we chose to use constant temperatures here as it provides a baseline for
186 characterizing thermal responses and because measuring all possible combinations of
187 temperature mean and variability would have been intractable.

188 The experiment began with larvae emerging 48 h after egg paper submersion (*i.e.*,
189 approximately 1-day old larvae). For each individual, we measured the following traits: larval
190 survival, larval development rate, pupal survival, pupal development rate, adult lifespan, and
191 wing length (a proxy for fecundity; see *Methods: Measuring wing length as a proxy for*
192 *fecundity*). We intentionally included more larvae from each population in the higher
193 temperature treatments as we expected greater mortality at these temperatures based on pilot
194 experiments. We visually inspected each individual on a daily basis, recording life stage
195 transitions and deaths, and moving individuals into the appropriate housing for the given life
196 stage. We maintained larvae in plastic containers in groups of five with approximately 100 mL of
197 water and 4 mg of larval food, in accordance with *Aedes* rearing protocols that promote high
198 larval survivorship in the absence of other factors^{31,32}. We maintained pupae individually in glass
199 vials with approximately 5 mL deionized water. Upon eclosion, we transferred adults to
200 individual 4 oz plastic specimen cups with one 10% sugar-soaked cotton ball and observed each
201 individual until death. To estimate fecundity of individuals that died as adults, we removed and
202 measured the length of their left wing, a commonly used proxy^{33,34} (see *Methods: Measuring*
203 *wing length as a proxy for fecundity*). Any larva that was alive but had not pupated by September
204 28, 2022 (*i.e.*, 82-84 days after larval emergence) was counted as survived for the larval survival
205 trait and considered to be in diapause.

206



207
 208 **Figure 1.** Sample collection locations and experimental design used to measure mosquito
 209 thermal performance. Ten populations were collected as larvae from tree holes across the
 210 Western U.S. and reared in the lab under common conditions for one generation. The resulting
 211 larvae from each population were randomly designated into one of six temperature treatments.
 212 The total number of larvae assigned to each treatment is noted above (n_{total}) as is the range of
 213 larvae from each population (n_{pop}); Supplemental Table S2 indicates the full breakdown of
 214 larvae per population and treatment). Individuals were checked daily for life stage transitions
 215 (e.g., larvae to pupae, pupae to adult) or death. Map colors denote the average maximum annual
 216 temperature (\square) from 1991 – 2020 from PRISM data. Supplemental Figure S1 shows the
 217 average minimum and mean temperature across this same extent. Population metadata, including
 218 full site names, latitude, longitude, and elevation are provided in Supplemental Table S1.

219
 220 *Measuring wing length as a proxy for fecundity*

221 To estimate individual fecundity, we measured the wing length of each individual used in the
 222 experiment. While wing length is an imperfect measure of fecundity, it is widely used in the
 223 literature and has been validated for several mosquito species (e.g., *Anopheles arabiensis*³³,
 224 *Anopheles gambiae*³⁴, *Aedes albopictus*^{35–37}, *Aedes geniculatus*³⁸) in addition to *Aedes*
 225 *sierrensis*³⁹. Further, using this proxy enabled us to obtain both a lifespan and estimate of
 226 reproductive output for each individual used in the experiment, whereas individually blood-
 227 feeding hundreds of mosquitoes held inside incubators would have been intractable. To measure
 228 wing length, we removed and photographed the left wing mounted on a microscope slide with a
 229 1 mm graticule. We then used ImageJ to measure the wing length as the distance from the alular
 230 notch to the tip of the wing margin excluding the fringe scales, using the 1 mm graticule for
 231 calibration⁴⁰ (Supplemental Figure S2). We then used the relationship between *Ae. sierrensis*
 232 female wing length and the number of eggs laid in the first clutch established in Washburn et al.

233 1989³⁹ (*i.e.*, 51.33 x female wing length (mm) - 87.96). We validated this relationship in the lab
234 using a separate, smaller number of individuals from our experimental populations (see
235 Supplemental Methods; Supplemental Figure S3).

236

237 *Fitness estimation*

238 We estimated an individual-level mosquito fitness proxy—here defined as a measure of
239 individual reproductive output through the first gonotrophic cycle—as survival to reproductive
240 maturity multiplied by estimated fecundity in the first gonotrophic cycle. For survival to
241 reproductive maturity, we considered whether an individual survived to adulthood and achieved
242 an adult lifespan of 10 days at 24 °C, 11 days at 17 °C, or 17 days at 13 °C. These lifespans
243 represent the minimum number of days from adult eclosion to egg-laying at a given temperature,
244 as observed in the validation experiment (*Supplemental Methods: Determining age at*
245 *reproductive maturity*). As no individuals eclosed at 5 °C and no individuals survived longer than
246 one day at 32 °C, all individuals at these two temperature treatments were estimated to have zero
247 fitness. For estimated fecundity, we used the wing length approximation described above
248 (*Methods: Measuring wing length as a proxy for fecundity*). As these estimates were made for
249 both males and females, we multiplied the estimated fecundity of a given adult by the proportion
250 of females from that population and temperature treatment.

251

252 *Characterizing the source thermal environment*

253 We characterized the source thermal environment of each population using climate data from
254 PRISM, which we accessed and analyzed using Google Earth Engine⁴¹. PRISM provides gridded
255 climate data at a 4 km resolution by downscaling data from a network of monitoring stations⁴².
256 We used either daily or monthly temperature data from 2000 – 2020 to calculate key variables
257 capturing temperature means, variations, and extremes. We specifically sought to include only
258 biologically meaningful temperature variables, such as those previously associated with thermal
259 tolerance in ectotherms⁴³, rather than many possible characterizations of climate (*e.g.*, all 19
260 WorldClim bioclimatic variables). These variables included annual mean temperature, mean
261 temperature in January – March (the period when eggs typically exit diapause and hatch as
262 larvae), seasonal variation in temperature (defined as the difference between the mean warmest
263 month temperature and the mean coolest month temperature), average warm-season maximum
264 (defined as the mean daily maxima in the Spring and Summer), and the number of days where
265 maximum temperatures exceeded 35 °C (the highest upper thermal limit for any trait estimated
266 from our experimental data) excluding periods of potential dormancy (*e.g.*, August – October).

267 Variables were calculated at a 1 km buffer around the sampled tree hole for each
268 population, approximating the geographic range of an individual mosquito. We investigated
269 Pearson’s correlations between these temperature variables and select thermal performance
270 parameters and traits (*i.e.*, those with significant between-population variation).

271 While the above estimates of source environmental temperature likely capture the
272 thermal conditions for populations at a broad spatial scale, they may not reflect the exact

273 temperatures within a given tree hole. We sought to directly measure tree hole temperatures for
274 each population by placing iButton temperature loggers (DS1921G, manufactured by Maxim
275 Integrated, San Jose, California) in each sampled tree hole at the time of location; however, only
276 two iButtons were recovered the following year. For these two tree holes, we compare the direct
277 temperature measurements made using the iButtons to the estimates from the PRISM data
278 described above.

279 To qualitatively understand how populations' estimated upper thermal limits compared to
280 source environmental conditions, we also calculated the number of days exceeding 31.6°C during
281 the adult activity period (e.g., March –July), as this was the estimated upper thermal limit for
282 adult lifespan (the lowest limit for any trait). However, we did not investigate correlations
283 between this environment variable and thermal performance characteristics to minimize multiple
284 testing. The *Ae. sierrensis* dormancy and adult activity windows described above were informed
285 by prior research in this system^{39,44}, as well as extensive *Ae. sierrensis* surveillance data available
286 from VectorSurv (<https://gateway.vectorsurv.org>). Specifically, we examined variation in
287 trapped adult abundance across the year using surveillance data from 2000 – 2020 for the trap
288 closest to each of our collection sites (Supplemental Figure S4).

289

290 *Analysis: Fitting thermal response curves*

291 To estimate the thermal limits and performance characteristics of each trait and population, we
292 fit thermal response curves to the experimental data using a Bayesian approach following
293 methods described in detail in Shocket et al. 2020⁴⁵. We first visually inspected the temperature-
294 performance data to determine the most appropriate functional form of the thermal response for
295 each trait. Consistent with prior work, we used quadratic fits truncated to a maximum of 1 for
296 larval and pupal survival, quadratic fits for adult lifespan, and Brière fits for larval and pupal
297 development rate and fitness^{45,46} (Supplemental Table S3).

298 We fit a first set of Bayesian models for each combination of trait and population across
299 temperatures using uniform priors for the thermal limit parameters bounded by biologically
300 plausible temperature cut-offs as in prior studies^{11,45–48} (i.e., trait performance was set to zero
301 below 0°C and above 40–45°C depending on the trait; Supplemental Table S3). For larval and
302 pupal development rate, adult lifespan, and fitness, we modeled the observed data as normally
303 distributed with the mean predicted by the thermal response function at that temperature and the
304 standard deviation, σ , as a gamma distributed parameter, $\frac{1}{\sigma^2}$, with shape parameter $\alpha = 0$ and rate
305 parameter $\beta = 1000$. For larval and pupal survival probabilities, we modeled the observed data as
306 binomially distributed with the probability and number of trials based on the proportional
307 survival and sample size for that temperature – population combination. We truncated thermal
308 response functions at zero for all traits, as well as at one for survival probability traits. We fit
309 models using Markov Chain Monte Carlo (MCMC) sampling, which uses simulation to
310 approximate the posterior distribution, using the 'R2jags' package⁴⁹. For each thermal response,
311 we ran three independent chains with a 5,000-iteration burn-in, and thinned the chains by saving
312 every eighth iteration. This fitting process produced 7,500 values in the posterior distribution for

313 each parameter of the thermal response function (*i.e.*, T_{\min} , T_{\max} and q) and enabled us to
314 calculate additional derived quantities for each trait and population including the maximum trait
315 performance value (P_{\max}), the temperature at maximum performance (T_{opt}), and the temperature
316 range where performance is at least 50% of the maximum (T_{breadth} ; See Supplemental Figure S5
317 for theoretical thermal performance curve). We refer to the above fitting process as our ‘low
318 information’ model specification.

319 To reduce the uncertainty in our parameter estimates, we then fit a second set of
320 models—the main models presented in the text—using informative priors generated using a two-
321 step process. In the first step, we specified low information priors as described above for each
322 population and trait but using only the temperature-performance data from the other nine
323 populations (*i.e.*, a ‘leave-one-out’ approach⁴⁵). We fit a Gamma probability distribution to the
324 posterior distributions of each thermal response parameter using the ‘MASS’ package⁵⁰. We then
325 used these hyperparameters as informative priors in a second round of model fitting. To ensure
326 the hyperparameters did not have an outsized influence on the resulting posterior distributions,
327 we increased the variance of the priors through multiplication by a constant k , set at 0.1 or 0.01,
328 depending on the trait (Supplemental Table S3). The parameter estimates from this ‘informative’
329 model specification are presented as the main results in the text but did not differ qualitatively
330 from those made through the ‘low information’ model specification presented in the supplement.
331 When investigating variation in thermal performance parameters, we interpreted non-overlapping
332 credible intervals as biologically meaningful and statistically supported differences between
333 populations and/or traits^{51–55}. It is worth noting that the leave-one-out informative prior approach
334 biases our thermal performance curve fits to be more similar across populations, making the
335 resulting estimates of differences among populations conservative. On the other hand, this
336 approach has the advantage of realistically constraining uncertainty, for example in cases where
337 a trait was poorly quantified at a given temperature (*i.e.*, few individuals in a given population
338 survived to the relevant life stage).

339

340 **Results**

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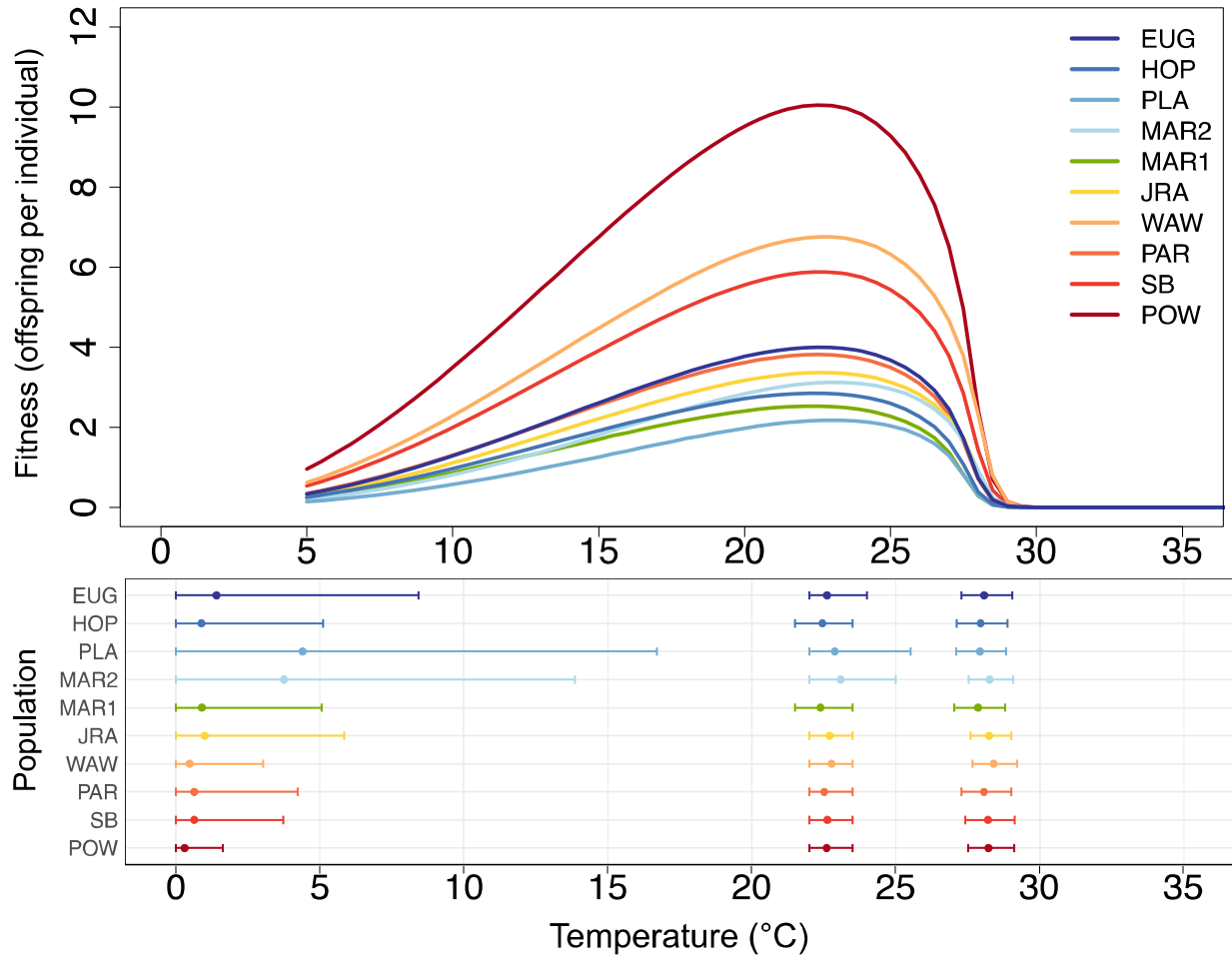
342 *How much does thermal tolerance vary between populations across the species range?*

343 We investigated variation in mosquito thermal performance between 10 populations across the
344 species range. For each population, we characterized the thermal performance of life history
345 traits constituting fitness by fitting thermal response curves (Figure 2, top panel) to our
346 experimental data and estimating the thermal limits and thermal optima (Figure 2, bottom panel).

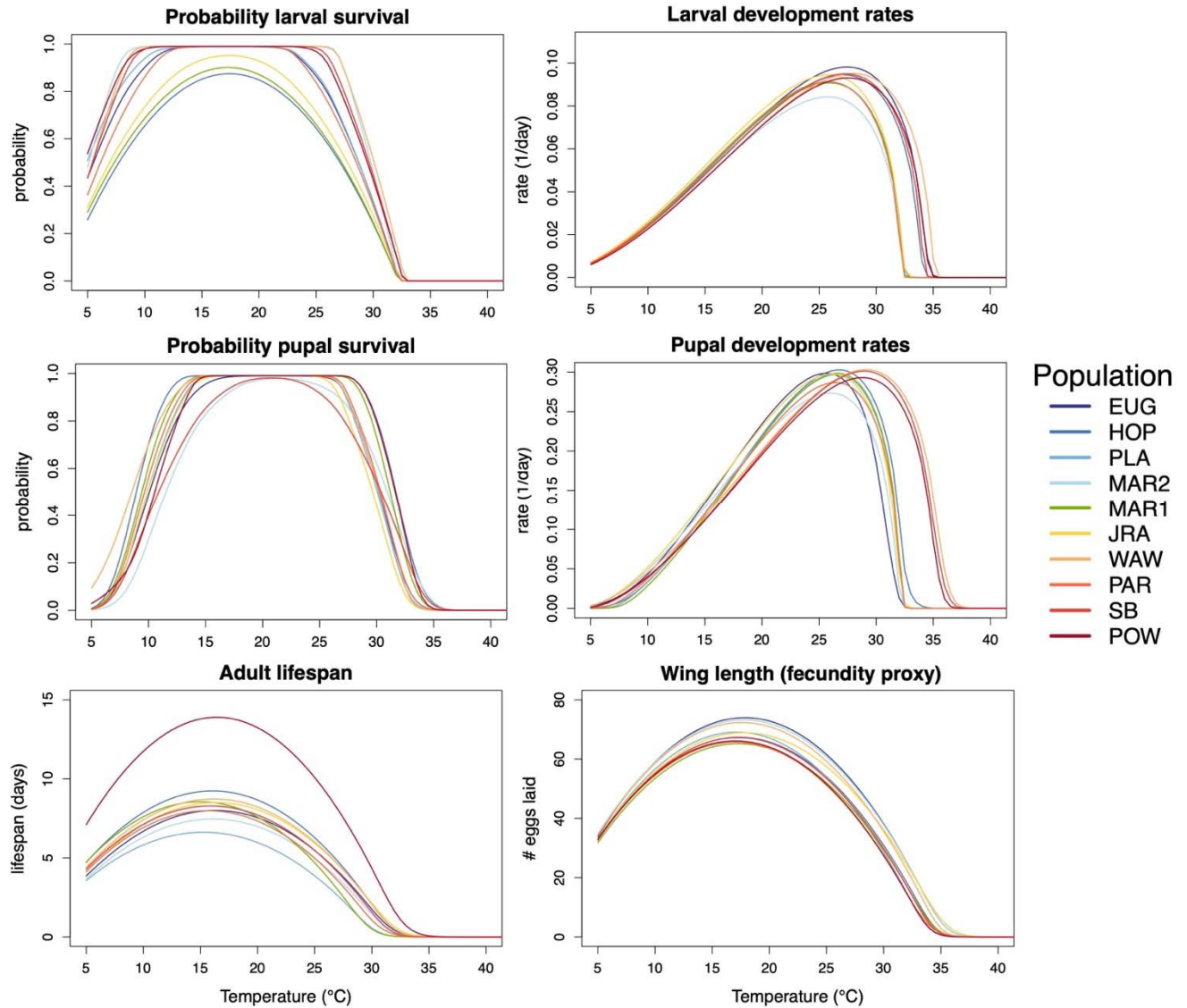
347 For our fitness proxy, we found very little variation in thermal tolerance between
348 populations (Figure 2). Specifically, both upper thermal limits and thermal optima varied by $<1^\circ\text{C}$
349 across all populations, ranging from $27.8 - 28.4^\circ\text{C}$ and $22.4 - 23.1^\circ\text{C}$, respectively. Further, the
350 95% credible intervals for these parameters overlapped for all populations, indicating non-
351 significant differences between populations. Populations displayed greater, but non-significant,
352 variation in their lower thermal limits for fitness, ranging from $0.3 - 4.6^\circ\text{C}$. These results were

353 highly similar when using the low information model specification (Supplemental Figures S8-9).
354 While it was not the focus of this study, we did also find that populations varied in maximum
355 fitness (P_{\max})—when averaging across temperature treatments, population’ maximum fitness
356 ranged from an estimated 2.2 – 10.1 offspring per individual (Supplemental Figure S6). We did
357 not detect between-population variation in the thermal breadth of fitness (Supplemental Figure
358 S6), nor any consistent correlations between fitness thermal performance characteristics (*i.e.*,
359 between P_{\max} and T_{breadth} or between P_{\max} and T_{opt}) among populations. These analyses and
360 results are discussed further in the Supplemental Results.

361 As with fitness, we found minimal variation in thermal tolerance between populations for
362 most individual life history traits (Figures 3-4). In particular, for all life history traits, both upper
363 and lower thermal limits varied by $<3^{\circ}\text{C}$ across populations (Figure 4, Supplemental Figure
364 S17). Similarly, thermal optima varied by $<1.5^{\circ}\text{C}$ for all traits except larval and pupal survival,
365 for which our estimates had the greatest uncertainty (partly due to high juvenile survivorship
366 across the intermediate temperature treatments). Variation between populations was non-
367 significant (*i.e.*, 95% credible intervals overlapped for all populations) for nearly all life history
368 traits and thermal performance parameters, with three exceptions: the upper thermal limits (T_{\max})
369 of larval and pupal development rates, and the thermal optima (T_{opt}) of pupal development rates.
370 Upper thermal limits for larval and pupal development rates each varied by 1.6°C across
371 populations ($33.3 - 34.9^{\circ}\text{C}$ and $32.1 - 33.7^{\circ}\text{C}$, respectively), while the thermal optima of pupal
372 development rate varied by 1.4°C ($26.3 - 27.7^{\circ}\text{C}$).
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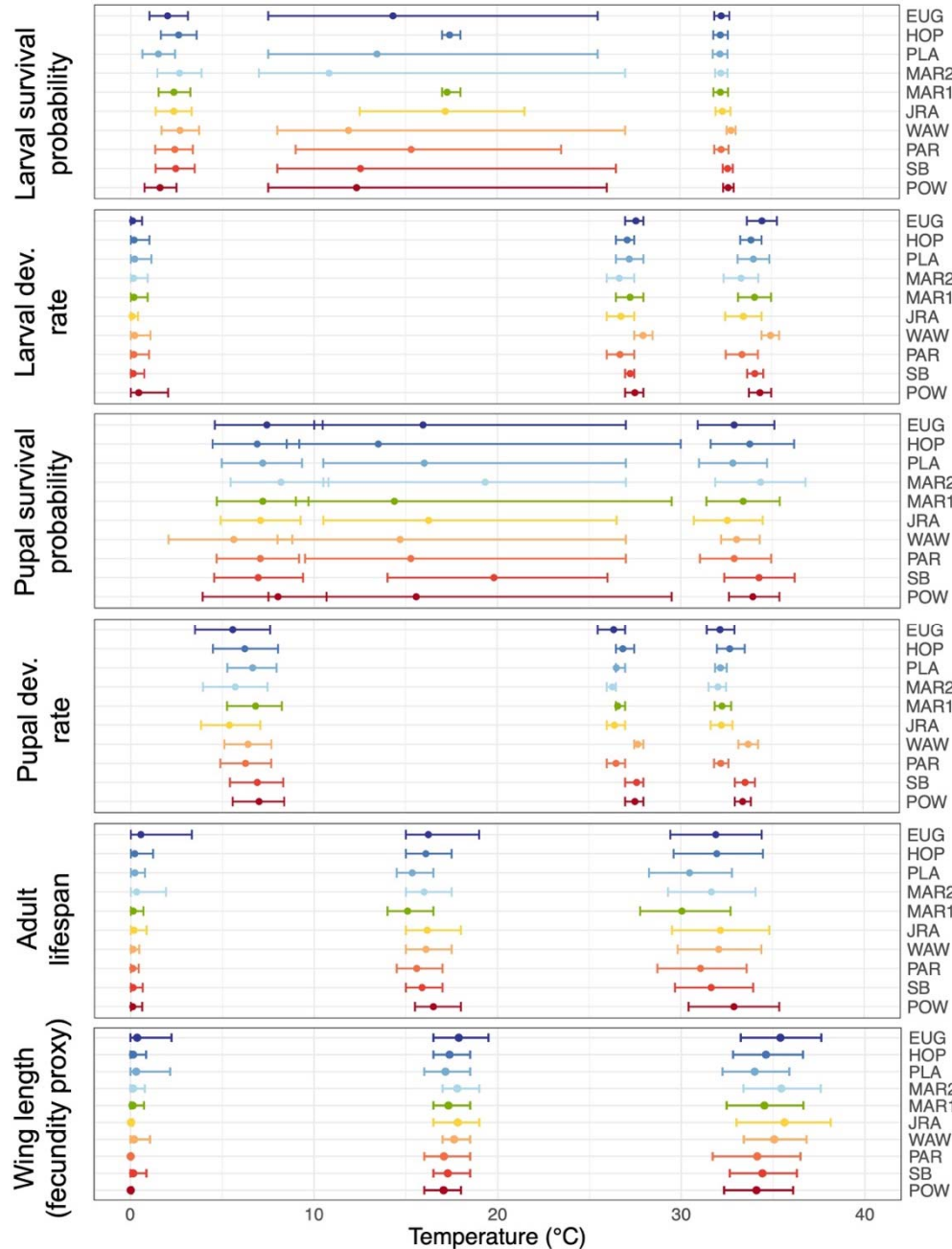


374
 375 **Figure 2. Populations vary minimally in their thermal limits and optima for fitness.** In the
 376 top panel, each line denotes the mean thermal performance of our fitness proxy for one
 377 population. In the bottom panel, points denote estimated thermal performance parameters for our
 378 fitness proxy for each population, including lower thermal limit (left), thermal optima (middle),
 379 and upper thermal limit (right). Error bars denote the 95% credible intervals for each parameter.
 380 In both panels, populations are colored and ordered by their latitude of collection from north
 381 (blue) to south (red); this color scheme and ordering is consistent across all figures in the paper.
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Figure 3. For most life history traits, thermal performance varies minimally between populations. Each curve denotes the average thermal performance for one population for a given trait. Populations are colored and ordered in the legend by their latitude of collection.



390

391 **Figure 4. Thermal limits and optima vary between life history traits, but minimally**

392 **between populations.** Lower thermal limits, thermal optima, and upper thermal limits for each

393 life history trait and population (left, middle, and right points and error bars in each panel,

394 respectively). Thermal performance parameter estimates are derived from the thermal

395 performance curves for traits for which the means are depicted in Figure 3. Points and error bars

396 denote the mean and 95% credible intervals for each parameter, respectively. Populations (listed

397 on the right) are colored and ordered by latitude of collection. Units of development rates and

398 lifespan are 1/days and days, respectively. Note that survival probability curves that are truncated

399 at one have very uncertain optimal temperatures because a wide range of temperatures have

400 similarly high survival probability.

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Is variation in thermal performance correlated with the source thermal environment?

To assess evidence of local thermal adaptation, we investigated the relationship between the source thermal environment (Table 1) and experimentally measured thermal performance parameters, using only the parameters with biologically significant between-population variation (*i.e.*, those where populations had non-overlapping 95% credible intervals). This included the upper thermal limits (T_{\max}) of larval and pupal development rates, and the thermal optima (T_{opt}) of pupal development rates.

We found several correlations that reflected patterns of local thermal adaptation. In particular, we found that T_{\max} and T_{opt} of pupal development were positively correlated with annual mean temperature, maximum daily temperatures in the Spring and Summer, and the number of days exceeding 35°C (r : 0.64 – 0.71; Figure 5). Together, this is consistent with local thermal adaptation of pupal development rate to high temperatures. By contrast, T_{\max} of larval development rate was not strongly correlated with any source temperature variable. We note that these reported correlations are only statistically significant ($p < 0.05$) prior to adjustment for multiple comparisons, the necessity of which is debated when making only specific, biologically meaningful comparisons (as we have done here) rather than all possible comparisons^{58,59}. The majority of the above correlations remained significant after removing ‘POW’ (Supplemental Table S6), the lowest latitude population, indicating that our findings of thermal adaptation are not solely driven by this population.

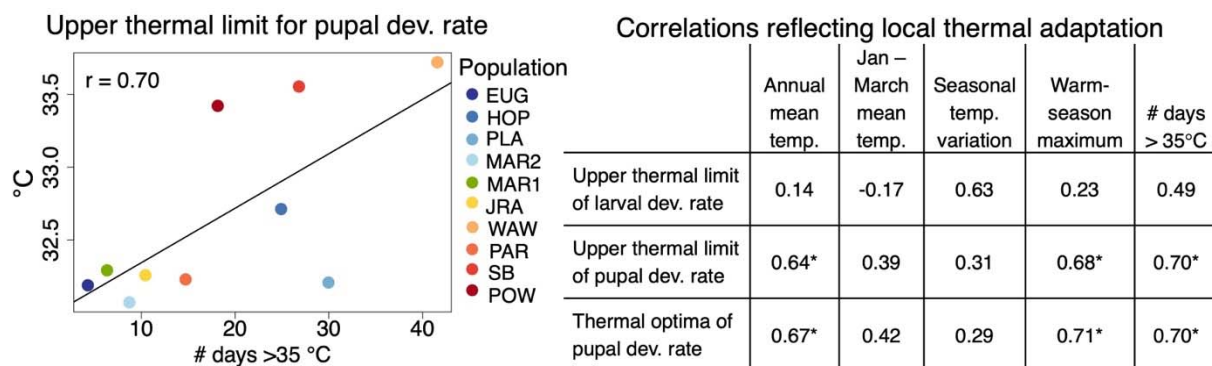
We also found that maximum fitness (P_{\max}), which varied significantly between populations, was positively correlated with annual mean temperature ($r = 0.66$; no correlations with other temperature variables were statistically significant). This result that populations from warmer climates generally have higher maximum fitness has frequently been found in other ectotherms^{60–62}, but does not necessarily reflect local thermal adaptation, in which peak fitness is expected to occur at conditions most similar to the source environment.

429 **Table 1.** Thermal characteristics of the source environment for each population listed in order of
 430 decreasing latitude (*i.e.*, north to south). Values represent averages from 2000 – 2020, calculated
 431 from PRISM climate data at a 1 km buffer around the sampled tree hole. Seasonal temperature
 432 variation is defined as the difference between the mean warmest month temperature and the
 433 mean coolest month temperature. Warm-season maximum is defined as the mean daily maxima
 434 in the Spring and Summer. The # days > 35 or 31.6°C refer to the average number of days where
 435 the maximum temperatures exceeded the stated threshold, either across the year, or when
 436 considering only non-dormant periods (January – July) or adult activity periods (March – July).
 437 See Supplemental Figure S18 for correlations between temperatures variables and Supplemental
 438 Figure S19 for comparisons between the PRISM and iButton temperature estimates for the ‘SB’
 439 and ‘POW’ populations.

Pop.	Annual mean temp. (°C)	Jan – March mean temp. (°C)	Seasonal temp. variation (°C)	Warm-season maximum (°C)	# days > 35°C		# days > 31.6°C	
					Across year	Jan. - July	Across year	March - July
EUG	11.48	6.33	16.01	22.69	4.29	2.23	18.33	9.45
HOP	14.57	9.41	15.38	26.77	24.90	11.95	57.67	28.863
PLA	16.08	9.91	18.08	28.04	29.95	17.00	76.24	40.27
MAR2	14.24	10.31	11.62	24.77	8.71	4.18	25.95	12.81
MAR1	14.04	10.18	11.29	24.41	6.33	3.05	20.86	10.36
JRA	15.45	11.49	11.61	25.29	10.43	4.77	27.14	13.54
WAW	15.83	8.73	20.51	28.19	41.57	20.95	87.76	44.18
PAR	14.40	9.88	13.26	26.52	14.71	5.77	46.67	22.73
SB	16.44	11.82	13.51	27.86	26.81	11.77	59.95	29.72
POW	18.75	15.08	11.65	27.77	18.14	5.09	50.62	21.09

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444 **Figure 5. Evidence of local thermal adaptation.** Correlations between the source thermal
 445 environment and population thermal performance provide evidence of local thermal adaptation
 446 (right). Statistically significant Pearson’s correlations (r ; $p < 0.05$) are denoted with (*). Note that
 447 correlations were only examined for traits with significant-between population variation. The
 448 relationship between upper thermal limits for pupal development rate and the number of days

449 with temperatures exceeding 35°C (one of the significant correlations noted in the table) is
450 visualized in the plot on the left.

451
452 *At present, how often do environmental temperatures exceed mosquito populations' upper*
453 *thermal limits?*

454 We found that for all populations, temperatures in the surrounding environment already exceed
455 our estimated upper thermal limits. In particular, the number of days per year with temperatures
456 exceeding 35°C—above the highest upper thermal limit we estimated for any life history trait—
457 ranged from 2 to 20 days (Table 1). This metric specifically excluded times of the year when *Ae.*
458 *sierrensis* populations are likely in dormancy (*e.g.*, August – October) – if all months were
459 included, there were an average of 4 to 42 days exceeding this threshold. Similarly, the number
460 of days exceeding 31.6°C—the lowest estimated upper thermal limit (adult lifespan)—ranged
461 from 9 to 40 days during adult activity season (*e.g.*, March – July) or 18 to 88 days across the
462 entire year.

463 The above estimates are based on PRISM climate data, which captures air temperature in
464 the broader surrounding environment, but not necessarily the precise temperature experienced in
465 a given tree hole. For two populations, we were able to record temperatures within the tree hole
466 for approximately one year following larval collection. We found that these direct measurements
467 were strongly correlated with temperature estimates from the PRISM climate data ($r = 0.91, 0.87$
468 for daily temperature estimates for the SB and POW populations, respectively; Supplemental
469 Figure S19). For these populations, the iButton recorded daily temperatures that were, on
470 average, 0.70°C higher (SB) or 3.0°C lower (POW) than the PRISM estimates. In both locations,
471 tree hole temperatures exceeded 31.6°C on several days (Supplemental Figure S19), indicating
472 that populations are exposed to temperatures above their estimated upper thermal limits for adult
473 lifespan even within this microhabitat.

474
475
476

477 Discussion

478
479 In one of the largest-ranging studies of standing variation in mosquito thermal tolerance to date,
480 we found limited evidence of variation between populations in the thermal responses of fitness
481 and life history traits. Specifically, in our common garden experiment using ten *Aedes sierrensis*
482 populations spanning over 1,200 km, we found the upper thermal limits and thermal optima for
483 fitness each varied by $<1^\circ\text{C}$ across all populations ($27.8 - 28.4^\circ\text{C}$ and $22.4 - 23.1^\circ\text{C}$, respectively;
484 Figure 2). This level of variation in upper thermal limits across latitude (*i.e.*, 0.6°C across
485 populations spanning 10° of latitude) is large relative to previous studies in terrestrial ectotherms
486 (0.3°C per 10° latitude⁵⁷); however, it is considerably less than the level of variation in
487 environmental temperature across this range, and likely less than the extent of warming expected
488 in this region in coming decades⁶³.

489 Our finding of minimal variation in mosquito thermal tolerance across the species range
490 is consistent with prior findings in a broad range of ectotherm species^{64,65}. For taxa including
491 insects, arachnids, reptiles, and amphibians, upper thermal limits typically vary little across wide
492 climatic and latitudinal gradients^{57,65-67}, a pattern that has been suggested to reflect hard
493 evolutionary constraints on heat tolerance^{68,69}. Although the underlying mechanism remains
494 unclear, the evolution of heat tolerance may be limited by genetic constraints (*e.g.*, low
495 heritability) and/or biochemical constraints (*e.g.*, limits on enzyme stability at high
496 temperatures)^{64,70,71}. Alternatively, this pattern could be driven by behavioral strategies enabling
497 populations to experience and adapt to similar thermal regimes across their range⁷², and/or trade-
498 offs in adapting to temperature versus other abiotic or biotic selection pressures⁷³.

499 Despite generally limited variation in thermal tolerance between populations, we did
500 observe meaningful variation in the thermal responses of larval and pupal development rates
501 (Figure 3). For both traits, upper thermal limits varied significantly, and by approximately 1.6°C
502 across populations—over twice as large as the variation estimated in fitness upper thermal limits
503 in our study and five times the average across terrestrial ectotherms spanning a similar latitudinal
504 extent⁵⁷ (Supplemental Table S4). Further, for pupal development rate, we found that variation in
505 populations' thermal optima and upper thermal limits was strongly correlated with variation in
506 the source thermal environment. Specifically, populations from environments with higher mean
507 and extreme temperatures had higher thermal optima and limits for pupal development rate than
508 those from cooler source environments, providing clear evidence for local thermal adaptation in
509 this trait (Figure 5).

510 That thermal adaptation was observed specifically in pupal development rate may be due
511 to the seasonal ecology of *Ae. sierrensis* making the pupal life stage the most vulnerable to high
512 temperatures. In particular, *Ae. sierrensis* eggs and larvae undergo a period of dormancy and are
513 primarily active earlier in the season, which may buffer these life stages from high temperature
514 extremes, while adults may avoid high temperatures through movement to cooler
515 microhabitats^{24,39}. Conversely, pupae have limited capacity for movement, no period of
516 dormancy, and typically begin development in the spring, which can have highly variable

517 thermal conditions across years and include high temperature extremes. This life history trait
518 may thus experience the strongest thermal selection pressure given the exposure to thermal stress
519 and a lack of other coping strategies. By measuring the thermal performance of traits across the
520 species life cycle, and using many populations from across a wide thermal gradient, we were
521 able to detect this specific evidence of thermal adaptation, which has not been clearly identified
522 in prior investigations of thermal adaptation in other mosquito species^{17,18,20}.

523 Despite this evidence of local thermal adaptation, the potential for further evolutionary
524 adaptation to warming could be limited. In addition to the minimal variation observed in upper
525 thermal limits for most traits, we found that temperatures at all source environments already
526 exceed our estimated upper thermal limits (Table 1). In particular, environmental temperatures at
527 each of our collection sites were at or above 35°C—exceeding the highest upper thermal limit we
528 estimated for any trait—for an average of 2 to 20 days out of the potential *Ae. sierrensis* activity
529 season (January - July). Similarly, environmental temperatures exceeded 31.6°C—the lowest
530 upper thermal limit across measured life history traits (adult lifespan, Figure 4)—for 9 to 40 days
531 during this period. Thus, populations may already be exposed to temperatures beyond their
532 estimated upper thermal limits; however, the extent to which this indicates climate vulnerability
533 depends on the time scales over which these high temperatures occur. In particular, short-term
534 thermal extremes (*e.g.*, one to several hours) that are followed by cooler temperatures could be
535 tolerated through heat stress repair, as has been found to occur during night-time in other
536 ectotherm species⁷⁴. As our experiments involved constant-temperature exposure, we were
537 unable to test whether such repair mechanisms could enable higher thermal tolerance –
538 incorporating diurnal temperature variation is an important next step for future experiments. In
539 addition to short-term heat repair, other strategies besides evolutionary adaptation, such as
540 seasonal life cycles and microhabitat selection may be important for sustaining *Ae. sierrensis*
541 under rapid climate warming. Accordingly, the majority of days exceeding the 35°C and 31.6°C
542 thresholds at our collection sites occurred after July, when most individuals in the population are
543 likely in the dormant egg stage (Supplementary Figure S3). Further, the tree hole microhabitat in
544 which *Ae. sierrensis* completes most of its life cycle may be cooler than the surrounding
545 environment, further buffering individuals from thermal extremes (although we found this was
546 not consistently the case; Supplementary Figure S18).

547 Phenological and behavioral strategies for mitigating thermal danger may be similarly
548 important for other mosquito and ectotherm species to persist under ongoing climate
549 warming^{72,75,76}. For example, *An. gambiae* in the Sahel have been shown to persist during the
550 arid summers by entering a prolonged period of dormancy⁷⁷, and winter dormancy responses are
551 widespread among mosquito species, likely facilitating their geographic expansion^{78,79}.
552 Similarly, biting activity in several mosquito species has been found to shift during warmer
553 months from dusk to late at night, although this was not conclusively linked to temperature.
554 Behavioral avoidance of high temperatures (typically >30°C) has been documented in adult
555 *Aedes*, *Anopheles*, and *Culex* spp. under lab conditions^{82–84}, and some evidence of preference for
556 cooler, shaded oviposition sites in warm climates has been found in field settings^{85,86}. These

557 types of strategies can have a large impact on buffering individuals from thermal stress^{72,75,87}, but
558 may dampen selection for greater thermal tolerance, further decreasing the likelihood of
559 evolutionary adaptation (termed the ‘Bogert effect’⁸⁸). Identifying the extent of behavioral
560 thermoregulation and temperature-driven changes in phenology in natural settings, and their
561 potential to enable mosquito persistence under climate warming are important directions for
562 future research.

563 Our experiment focused on the impacts of constant temperatures on mosquito trait
564 performance—an important first step in characterizing thermal tolerance for a given species.
565 However, changes in temperature fluctuations and short-term thermal extremes are key
566 components of climate warming projections and can have a large impact on mosquito life
567 histories^{89–94}. In particular, coping with large fluctuations in temperature and/or acute thermal
568 extremes may require a different set of physiological or behavioral strategies than coping with
569 constant warm temperatures^{95–97}. Thus, patterns of mosquito thermal adaptation to these aspects
570 of temperature could differ from those estimated here. However, to our knowledge, no studies
571 have yet measured variation in mosquito responses to mean, fluctuating, and extreme
572 temperatures between populations. Prior studies in other ectotherm species have tested whether
573 thermal performance under fluctuating temperatures can be predicted qualitatively from thermal
574 performance curves estimated at constant temperatures, finding mixed results^{98,99}.
575 Experimentally testing this approach in mosquitoes and estimating mosquito performance under
576 thermal regimes that reflect natural conditions using populations from across the species range
577 are important future directions.

578

579 **Authors’ Contributions**

580 L.I.C. and E.A.M. conceived of and designed the project. L.I.C., J.E.F, and K.P.L. performed the
581 field collection and laboratory rearing. L.I.C., J.E.F., K.P.L., and A.S.L. conducted the
582 experiment. L.I.C. conducted the analyses and drafted the manuscript. All authors revised the
583 manuscript and read and approved the final version.

584

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602

603 **Conflict of Interest**

604 The authors declare that they have no conflicts of interest.

605

606 **Data accessibility**

607 All data and code used in this project are publicly available on GitHub:

608 <https://github.com/lcouper/MosquitoThermalAdaptation>

609

610 **References**

- 611 1. Mordecai, E. A. *et al.* Thermal biology of mosquito-borne disease. *Ecol. Lett.* **22**, 1690–1708 (2019).
- 612 2. Deutsch, C. A. *et al.* Impacts of climate warming on terrestrial ectotherms across latitude. *Proc. Natl.*
613 *Acad. Sci.* **105**, 6668–6672 (2008).
- 614 3. Ryan, S. J., Carlson, C. J., Mordecai, E. A. & Johnson, L. R. Global expansion and redistribution of
615 *Aedes*-borne virus transmission risk with climate change. *PLoS Negl. Trop. Dis.* **13**, e0007213
616 (2019).
- 617 4. Servadio, J. L., Rosenthal, S. R., Carlson, L. & Bauer, C. Climate patterns and mosquito-borne
618 disease outbreaks in South and Southeast Asia. *J. Infect. Public Health* **11**, 566–571 (2018).
- 619 5. Lafferty, K. D. The ecology of climate change and infectious diseases. *Ecology* **90**, 888–900 (2009).
- 620 6. Gething, P. W. *et al.* Climate change and the global malaria recession. *Nature* **465**, 342–345 (2010).
- 621 7. Aitken, S. N., Yeaman, S., Holliday, J. A., Wang, T. & Curtis-McLane, S. Adaptation, migration or
622 extirpation: climate change outcomes for tree populations. *Evol. Appl.* **1**, 95–111 (2008).
- 623 8. Couper, L. I. *et al.* How will mosquitoes adapt to climate warming? *eLife* **10**, e69630 (2021).
- 624 9. Pawar, S. *et al.* Variation in temperature of peak trait performance will constrain adaptation of
625 arthropod populations to climatic warming. *bioRxiv* (2023) doi:10.1101/2023.01.18.524448.
- 626 10. Sternberg, E. D. & Thomas, M. B. Local adaptation to temperature and the implications for vector-
627 borne diseases. *Trends Parasitol.* **30**, 115–122 (2014).
- 628 11. Villena, O. C., Ryan, S. J., Murdock, C. C. & Johnson, L. R. Temperature impacts the environmental
629 suitability for malaria transmission by *Anopheles gambiae* and *Anopheles stephensi*. *Ecology* **103**,
630 e3685 (2022).
- 631 12. Oliveira, B. F., Yogo, W. I. G., Hahn, D. A., Yongxing, J. & Scheffers, B. R. Community-wide
632 seasonal shifts in thermal tolerances of mosquitoes. *Ecology* **102**, e03368 (2021).
- 633 13. Medley, K. A., Westby, K. M. & Jenkins, D. G. Rapid local adaptation to northern winters in the
634 invasive Asian tiger mosquito *Aedes albopictus*: A moving target. *J. Appl. Ecol.* **56**, 2518–2527
635 (2019).
- 636 14. Urbanski, J. *et al.* Rapid Adaptive Evolution of Photoperiodic Response during Invasion and Range
637 Expansion across a Climatic Gradient. *Am. Nat.* **179**, 490–500 (2012).
- 638 15. Fouet, C., Gray, E., Besansky, N. J. & Costantini, C. Adaptation to Aridity in the Malaria Mosquito
639 *Anopheles gambiae*: Chromosomal Inversion Polymorphism and Body Size Influence Resistance to
640 Desiccation. *PLOS ONE* **7**, e34841 (2012).
- 641 16. Gray, E. M., Rocca, K. A., Costantini, C. & Besansky, N. J. Inversion 2La is associated with
642 enhanced desiccation resistance in *Anopheles gambiae*. *Malar. J.* **8**, 215 (2009).

- 643 17. Ruybal, J. E., Kramer, L. D. & Kilpatrick, A. M. Geographic variation in the response of *Culex*
644 *pipiens* life history traits to temperature. *Parasit. Vectors* **9**, 116 (2016).
- 645 18. Reisen, W. Effect of Temperature on *Culex tarsalis* (Diptera: Culicidae) from the Coachella and San
646 Joaquin Valleys of California. *J. Med. Entomol.* **32**, 636–645 (1995).
- 647 19. Vorhees, A. S., Gray, E. M. & Bradley, T. J. Thermal Resistance and Performance Correlate with
648 Climate in Populations of a Widespread Mosquito. *Physiol. Biochem. Zool.* **86**, 73–81 (2013).
- 649 20. Chu, V. M. *et al.* Regional variation in life history traits and plastic responses to temperature of the
650 major malaria vector *Nyssonorhynchus darlingi* in Brazil. *Sci. Rep.* **9**, 5356 (2019).
- 651 21. Zani, P. A. *et al.* Geographic Variation in Tolerance of Transient Thermal Stress in the Mosquito
652 *Wyeomyia Smithii*. *Ecology* **86**, 1206–1211 (2005).
- 653 22. Orr, H. A. Fitness and its role in evolutionary genetics. *Nat. Rev. Genet.* **10**, 531–539 (2009).
- 654 23. Dobson, F. S., Murie, J. O. & Viblanc, V. A. Fitness Estimation for Ecological Studies: An
655 Evaluation in Columbian Ground Squirrels. *Front. Ecol. Evol.* **8**, (2020).
- 656 24. Hawley, W. A. The Effect of Larval Density on Adult Longevity of a Mosquito, *Aedes sierrensis*:
657 Epidemiological Consequences. *J. Anim. Ecol.* **54**, 955 (1985).
- 658 25. Garcia, R. & Pontig, G. Studies on the ecology of the treehole mosquito *Aedes sierrensis* (Ludlow).
659 *Calif. Mosq. Control Assoc. Proc.* (1972).
- 660 26. Ware-Gilmore, F., Novelo, M., Sgrò, C. M., Hall, M. D. & McGraw, E. A. Assessing the role of
661 family level variation and heat shock gene expression in the thermal stress response of the mosquito
662 *Aedes aegypti*. *Philos. Trans. R. Soc. B Biol. Sci.* **378**, 20220011 (2023).
- 663 27. Medley, K. A. Niche shifts during the global invasion of the Asian tiger mosquito, *Aedes albopictus*
664 Skuse (Culicidae), revealed by reciprocal distribution models. *Glob. Ecol. Biogeogr.* **19**, 122–133
665 (2010).
- 666 28. Sims, S. R. Larval Diapause in the Eastern Tree-Hole Mosquito, *Aedes triseriatus*1: Latitudinal
667 Variation in Induction and Intensity. *Ann. Entomol. Soc. Am.* **75**, 195–200 (1982).
- 668 29. Chaves, L. F., Reissen, N., White, G. S., Gordon, S. & Faraji, A. Trap Comparison for Surveillance
669 of the Western Tree Hole Mosquito, *Aedes sierrensis* (Diptera: Culicidae). *J. Insect Sci.* **20**, 3 (2020).
- 670 30. Verheyen, J., Tüzün, N. & Stoks, R. Using natural laboratories to study evolution to global warming:
671 contrasting altitudinal, latitudinal, and urbanization gradients. *Curr. Opin. Insect Sci.* **35**, 10–19
672 (2019).
- 673 31. Puggioli, A. *et al.* Development of *Aedes albopictus* (Diptera: Culicidae) Larvae Under Different
674 Laboratory Conditions. *J. Med. Entomol.* **54**, 142–149 (2017).
- 675 32. International Atomic Energy Agency. Guidelines for routine colony maintenance of *Aedes* mosquito
676 species. (2017).
- 677 33. Ameshewa, B. & Service, M. W. The relationship between female body size and survival rate of
678 the malaria vector *Anopheles arabiensis* in Ethiopia. *Med. Vet. Entomol.* **10**, 170–172 (1996).
- 679 34. Lyimo, E. O. & Takken, W. Effects of adult body size on fecundity and the pre-gravid rate of
680 *Anopheles gambiae* females in Tanzania. *Med. Vet. Entomol.* **7**, 328–332 (1993).
- 681 35. Blackmore, M. S. & Lord, C. C. The relationship between size and fecundity in *Aedes albopictus*. *J.*
682 *Vector Ecol.* **25**, 212–217 (2000).
- 683 36. Davis, T. J., Kline, D. L. & Kaufman, P. E. Assessment of *Aedes albopictus* (Skuse) (Diptera:
684 Culicidae) clutch size in wild and laboratory populations. *J. Vector Ecol.* **41**, 11–17 (2016).
- 685 37. Costanzo, K. S., Westby, K. M. & Medley, K. A. Genetic and environmental influences on the size-
686 fecundity relationship in *Aedes albopictus* (Diptera: Culicidae): Impacts on population growth
687 estimates? *PLOS ONE* **13**, e0201465 (2018).
- 688 38. Armbruster, P. & Hutchinson, R. A. Pupal Mass and Wing Length as Indicators of Fecundity in
689 *Aedes albopictus* and *Aedes geniculatus* (Diptera: Culicidae). *J. Med. Entomol.* **39**, 699–704 (2002).
- 690 39. Washburn, J. O., Anderson, J. R. & Mercer, D. R. Emergence Characteristics of *Aedes sierrensis*
691 (Diptera: Culicidae) from California Treeholes with Particular Reference to Parasite Loads. *J. Med.*
692 *Entomol.* **26**, 173–182 (1989).

- 693 40. Abramoff, M. D., Magalhães, P. J. & Ram, S. J. Image processing with ImageJ. *Biophotonics Int.* **11**,
694 36–42 (2004).
- 695 41. Gorelick, N. *et al.* Google Earth Engine: Planetary-scale geospatial analysis for everyone. *Remote*
696 *Sens. Environ.* **202**, 18–27 (2017).
- 697 42. Daly, C. *et al.* Physiographically sensitive mapping of climatological temperature and precipitation
698 across the conterminous United States. *Int. J. Climatol.* **28**, 2031–2064 (2008).
- 699 43. Sunday, J. *et al.* Thermal tolerance patterns across latitude and elevation. *Philos. Trans. R. Soc. B*
700 *Biol. Sci.* **374**, 20190036 (2019).
- 701 44. Jordan, R. G. & Bradshaw, W. E. Geographic Variation in the Photoperiodic Response of the
702 Western Tree-Hole Mosquito, *Aedes sierrensis*. *Ann. Entomol. Soc. Am.* **71**, 487–490 (1978).
- 703 45. Shocket, M. S. *et al.* Transmission of West Nile and five other temperate mosquito-borne viruses
704 peaks at temperatures between 23°C and 26°C. *eLife* **9**, e58511 (2020).
- 705 46. Mordecai, E. A. *et al.* Detecting the impact of temperature on transmission of Zika, dengue, and
706 chikungunya using mechanistic models. *PLoS Negl. Trop. Dis.* **11**, e0005568 (2017).
- 707 47. Johnson, L. R. *et al.* Understanding uncertainty in temperature effects on vector-borne disease: a
708 Bayesian approach. *Ecology* **96**, 203–213 (2015).
- 709 48. Tesla, B. *et al.* Temperature drives Zika virus transmission: evidence from empirical and
710 mathematical models. *Proc. R. Soc. B Biol. Sci.* (2018).
- 711 49. Su, Y. & Yajima, M. R2jags: A Package for Running jags from R. (2009).
- 712 50. Venables, W. & Ripley, B. *Modern Applied Statistics with S.* (Springer, 2002).
- 713 51. Hespanhol, L., Vallio, C. S., Costa, L. M. & Saragiotto, B. T. Understanding and interpreting
714 confidence and credible intervals around effect estimates. *Braz. J. Phys. Ther.* (2019).
- 715 52. Kelter, R. Bayesian alternatives to null hypothesis significance testing in biomedical research: a non-
716 technical introduction to Bayesian inference with JASP. *BMC Med. Res. Methodol.* **20**, 142 (2020).
- 717 53. Kunze, C., Luijckx, P., Jackson, A. L. & Donohue, I. Alternate patterns of temperature variation
718 bring about very different disease outcomes at different mean temperatures. *eLife* **11**, e72861 (2022).
- 719 54. Stefanovic, D. I. *et al.* Thermal stress and the heat shock response in embryonic and young of the
720 year juvenile lake whitefish. *Comp. Biochem. Physiol. A. Mol. Integr. Physiol.* **193**, 1–10 (2016).
- 721 55. Whitehouse, L. M. *et al.* Development of the embryonic heat shock response and the impact of
722 repeated thermal stress in early stage lake whitefish (*Coregonus clupeaformis*) embryos. *J. Therm.*
723 *Biol.* **69**, 294–301 (2017).
- 724 56. Kellermann, V. & van Heerwaarden, B. Terrestrial insects and climate change: adaptive responses in
725 key traits. *Physiol. Entomol.* **44**, 99–115 (2019).
- 726 57. Sunday, J. M., Bates, A. E. & Dulvy, N. K. Global analysis of thermal tolerance and latitude in
727 ectotherms. *Proc. R. Soc. B Biol. Sci.* **278**, 1823–1830 (2011).
- 728 58. Rothman, K. J. No Adjustments Are Needed for Multiple Comparisons. *Epidemiology* **1**, 43–46
729 (1990).
- 730 59. Frane, A. V. Planned Hypothesis Tests Are Not Necessarily Exempt From Multiplicity Adjustment.
731 *J. Res. Pract.* (2015).
- 732 60. Knies, J. L., Kingsolver, J. G. & Burch, C. L. Hotter Is Better and Broader: Thermal Sensitivity of
733 Fitness in a Population of Bacteriophages. *Am. Nat.* **173**, 419–430 (2009).
- 734 61. Angilletta, M. J. *Thermal Adaptation: A Theoretical and Empirical Synthesis.* (OUP Oxford, 2009).
- 735 62. Kontopoulos, D.-G. *et al.* Phytoplankton thermal responses adapt in the absence of hard
736 thermodynamic constraints. *Evolution* **74**, 775–790 (2020).
- 737 63. Walton, D. *et al.* Understanding Differences in California Climate Projections Produced by
738 Dynamical and Statistical Downscaling. *J. Geophys. Res. Atmospheres* **125**, e2020JD032812 (2020).
- 739 64. Tüzün, N. & Stoks, R. Evolution of geographic variation in thermal performance curves in the face of
740 climate change and implications for biotic interactions. *Curr. Opin. Insect Sci.* **29**, 78–84 (2018).
- 741 65. Klepsatel, P. *et al.* Variation in Thermal Performance and Reaction Norms Among Populations of
742 *Drosophila Melanogaster*. *Evolution* **67**, 3573–3587 (2013).

- 743 66. Addo-Bediako, A., Chown, S. L. & Gaston, K. J. Thermal tolerance, climatic variability and latitude.
744 *Proc. R. Soc. Lond. B Biol. Sci.* **267**, 739–745 (2000).
- 745 67. MacLean, H. J. *et al.* Evolution and plasticity of thermal performance: an analysis of variation in
746 thermal tolerance and fitness in 22 *Drosophila* species. *Philos. Trans. R. Soc. B Biol. Sci.* **374**,
747 20180548 (2019).
- 748 68. Bennett, J. M. *et al.* The evolution of critical thermal limits of life on Earth. *Nat. Commun.* **12**, 1198
749 (2021).
- 750 69. González-Tokman, D. *et al.* Insect responses to heat: physiological mechanisms, evolution and
751 ecological implications in a warming world. *Biol. Rev.* **95**, 802–821 (2020).
- 752 70. Angilletta, M. J., Wilson, R. S., Navas, C. A. & James, R. S. Tradeoffs and the evolution of thermal
753 reaction norms. *Trends Ecol. Evol.* **18**, 234–240 (2003).
- 754 71. Thermal adaptation revisited: How conserved are thermal traits of reptiles and amphibians? -
755 Bodensteiner - 2021 - Journal of Experimental Zoology Part A: Ecological and Integrative
756 Physiology - Wiley Online Library. <https://onlinelibrary.wiley.com/doi/full/10.1002/jez.2414>.
- 757 72. Sunday, J. M. *et al.* Thermal-safety margins and the necessity of thermoregulatory behavior across
758 latitude and elevation. *Proc. Natl. Acad. Sci.* **111**, 5610–5615 (2014).
- 759 73. Pörtner, H. O. *et al.* Tradeoffs in Thermal Adaptation: The Need for a Molecular to Ecological
760 Integration. *Physiol. Biochem. Zool.* **79**, 295–313 (2006).
- 761 74. Bai, C.-M., Ma, G., Cai, W.-Z. & Ma, C.-S. Independent and combined effects of daytime heat stress
762 and night-time recovery determine thermal performance. *Biol. Open* **8**, bio038141 (2019).
- 763 75. Kearney, M., Shine, R. & Porter, W. P. The potential for behavioral thermoregulation to buffer “cold-
764 blooded” animals against climate warming. *Proc. Natl. Acad. Sci.* **106**, 3835–3840 (2009).
- 765 76. Bradshaw, W. E., Fujiyama, S. & Holzapfel, C. M. Adaptation to the Thermal Climate of North
766 America by the Pitcher-Plant Mosquito, *Wyeomyia Smithii*. *Ecology* **81**, 1262–1272 (2000).
- 767 77. Lehmann, T. *et al.* Aestivation of the African Malaria Mosquito, *Anopheles gambiae* in the Sahel.
768 *Am. J. Trop. Med. Hyg.* **83**, 601–606 (2010).
- 769 78. Denlinger, D. L. & Armbruster, P. A. Mosquito Diapause. *Annu. Rev. Entomol.* **59**, 73–93 (2014).
- 770 79. Diniz, D. F. A., de Albuquerque, C. M. R., Oliva, L. O., de Melo-Santos, M. A. V. & Ayres, C. F. J.
771 Diapause and quiescence: dormancy mechanisms that contribute to the geographical expansion of
772 mosquitoes and their evolutionary success. *Parasit. Vectors* **10**, (2017).
- 773 80. Reisen, W. K. & Aslamkhan, M. Biting rhythms of some Pakistan mosquitoes (Diptera: Culicidae).
774 *Bull. Entomol. Res.* **68**, 313–330 (1978).
- 775 81. Voorham, J. Intra-population plasticity of *Anopheles darlingi*'s (Diptera, Culicidae) biting activity
776 patterns in the state of Amapá, Brazil. *Rev. Saúde Pública* **36**, 75–80 (2002).
- 777 82. Ziegler, R., Blanckenhorn, W. U., Mathis, A. & Verhulst, N. O. Video analysis of the locomotory
778 behaviour of *Aedes aegypti* and *Ae. japonicus* mosquitoes under different temperature regimes in a
779 laboratory setting. *J. Therm. Biol.* **105**, 103205 (2022).
- 780 83. Verhulst, N. O., Brendle, A., Blanckenhorn, W. U. & Mathis, A. Thermal preferences of subtropical
781 *Aedes aegypti* and temperate *Ae. japonicus* mosquitoes. *J. Therm. Biol.* **91**, 102637 (2020).
- 782 84. Blanford, S., Read, A. F. & Thomas, M. B. Thermal behaviour of *Anopheles stephensi* in response to
783 infection with malaria and fungal entomopathogens. *Malar. J.* **8**, (2009).
- 784 85. Vezzani, D. & Albicocco, A. P. The effect of shade on the container index and pupal productivity of
785 the mosquitoes *Aedes aegypti* and *Culex pipiens* breeding in artificial containers. *Med. Vet. Entomol.*
786 **23**, 78–84 (2009).
- 787 86. Barrera, R., Amador, M. & Clark, G. G. Ecological Factors Influencing *Aedes aegypti* (Diptera:
788 Culicidae) Productivity in Artificial Containers in Salinas, Puerto Rico. *J. Med. Entomol.* **43**, (2006).
- 789 87. Pinsky, M. L., Eikeset, A. M., McCauley, D. J., Payne, J. L. & Sunday, J. M. Greater vulnerability to
790 warming of marine versus terrestrial ectotherms. *Nature* **569**, 108–111 (2019).
- 791 88. Huey, R. B., Hertz, P. E. & Sinervo, B. Behavioral Drive versus Behavioral Inertia in Evolution: A
792 Null Model Approach. *Am. Nat.* **161**, 357–366 (2003).

- 793 89. Paaijmans, K. P. *et al.* Temperature variation makes ectotherms more sensitive to climate change.
794 *Glob. Change Biol.* **19**, 2373–2380 (2013).
- 795 90. Lambrechts, L. *et al.* Impact of daily temperature fluctuations on dengue virus transmission by *Aedes*
796 *aegypti*. *Proc. Natl. Acad. Sci.* **108**, 7460–7465 (2011).
- 797 91. Alto, B. W. *et al.* Diurnal Temperature Range and Chikungunya Virus Infection in Invasive Mosquito
798 Vectors. *J. Med. Entomol.* **55**, 217–224 (2018).
- 799 92. Paaijmans, K. P. *et al.* Influence of climate on malaria transmission depends on daily temperature
800 variation. *Proc. Natl. Acad. Sci.* **107**, 15135–15139 (2010).
- 801 93. Beck-Johnson, L. M. *et al.* The importance of temperature fluctuations in understanding mosquito
802 population dynamics and malaria risk. *R. Soc. Open Sci.* **4**, 160969 (2017).
- 803 94. Bradshaw, W. E. Thermoperiodism and the thermal environment of the pitcher-plant mosquito,
804 *Wyeomyia smithii*. *Oecologia* **46**, 13–17 (1980).
- 805 95. Hoffmann, A. A. Physiological climatic limits in *Drosophila*: patterns and implications. *J. Exp. Biol.*
806 **213**, 870–880 (2010).
- 807 96. Dowd, W. W., King, F. A. & Denny, M. W. Thermal variation, thermal extremes and the
808 physiological performance of individuals. *J. Exp. Biol.* **218**, 1956–1967 (2015).
- 809 97. Clusella-Trullas, S., Blackburn, T. M. & Chown, S. L. Climatic Predictors of Temperature
810 Performance Curve Parameters in Ectotherms Imply Complex Responses to Climate Change. *Am.*
811 *Nat.* **177**, 738–751 (2011).
- 812 98. Bernhardt, J. R., Sunday, J. M., Thompson, P. L. & O'Connor, M. I. Nonlinear averaging of thermal
813 experience predicts population growth rates in a thermally variable environment. *Proc. R. Soc. B*
814 *Biol. Sci.* **285**, 20181076 (2018).
- 815 99. Marshall, K. E. *et al.* Whole-organism responses to constant temperatures do not predict responses to
816 variable temperatures in the ecosystem engineer *Mytilus trossulus*. *Proc. R. Soc. B Biol. Sci.* **288**,
817 20202968 (2021).
- 818