

1 **Maximum movement performance, not activity levels or thermoregulatory indices,**
2 **affects survival in a free-ranging ectotherm**

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Abstract

1. Temperature profoundly influences the distribution and diversity of ectotherms, yet in natural settings, the trade-offs between environmental temperatures, behaviour, physiological function, and how they drive individual survival, remain poorly understood.
2. To address these gaps, we generated field-based thermal performance curves, using temperature-sensitive radio transmitters and accelerometers, to examine the links between thermal biology and survival in a widely distributed, well-studied, lizard: the Australian central bearded dragon (*Pogona vitticeps*). Using this approach, we were able to link the accuracy and effectiveness of thermoregulation in the wild to address if seasonal trade-offs of thermoregulatory behaviours affect individual performance and ultimately test if thermal performance and thermoregulatory behaviours predict survival.
3. Lizards adjusted their behaviour to maintain optimal body temperature with higher thermoregulatory precision when environmental costs to maintain body temperature were low, during spring and summer, but decreased that precision during winter when the costs of active thermoregulation were high.
4. Maximum movement performance in the field was a strong predictor of individual survival, regardless of sex, even though survival probabilities were higher in males than females. Specifically, higher maximum movement performance increased predation and, as a result, mortality risk. Conversely, survival was not related to activity levels or thermoregulatory indices.
5. These findings highlight the complex trade-offs that ectotherms must navigate to balance behavioural thermoregulation and survival, emphasising the importance of seasonal and sex-specific differences in making those trade-offs. Connecting accelerometer data to energy expenditure and to specific behaviours is a promising avenue for research into thermal ecology. Such data provides critical insights into how ectotherms can respond to climate warming.

49 **1. Introduction**

50 In meeting the competing demands on their time, animals must balance the costs and benefits
51 of their behaviours where these decisions have an influence on their survival and reproductive
52 success, that is, their fitness (Huey & Slatkin, 1976). The energy expended in undertaking
53 those behaviours needs to be weighed against energetic costs and fitness gain. Trade-offs are
54 inevitable, and manifest in such contexts as optimal foraging behaviour, investment in mating
55 displays and territorial defence and in migration. This intricate balance is underpinned by a
56 rich and diverse body of literature that spans decades and species (Boyd & Hoelzel, 2002;
57 Brown et al., 2018; Campos-Candela, 2018; Huey & Slatkin, 1976). Understanding these
58 trade-offs reveals the evolutionary forces that shape behaviour, especially as animals navigate
59 their environments to maximise fitness.

60 Ectotherms, those animals that maintain body temperatures within narrow limits while
61 active by drawing upon external thermal sources and sinks, have a unique set of trade-offs in
62 order to maximise fitness (Huey, 1982; Huey & Slatkin, 1976). Trade-offs that involve when
63 and when not to be active, shuttling between sun and shade, which compromises the ability to
64 forage or find mates, exposure to heat sources, but also predation risk are all optimised within
65 a narrow range of body temperatures (Angilletta et al., 2002; Porter et al., 1973; Van Damme
66 et al., 1991). However, these trade-offs can be energetically costly or increase predation risk
67 depending on the environment (Herczeg et al., 2008; Kearney et al., 2009; Sears & Angilletta,
68 2015). What remains unclear is how these decisions affect individual survival under natural
69 conditions. Ectotherms thus provide unique opportunities to examine the relationship between
70 animals and their environment in balancing the competing demands of survival and
71 reproduction to maximise their fitness.

72 Presumed links between body temperature and survival underpin the study of thermal
73 ecology in ectotherms. In particular, behavioural thermoregulation is assumed to enable
74 individuals to achieve optimal body temperature and hence maximise their physiological
75 performance (Angilletta, 2009). This, in turn, can provide performance advantages, such as
76 enhanced digestion efficiency or increased sprint speed, that may influence individual fitness
77 (Angilletta et al., 2002; Pearson & Warner, 2018). However, the benefits of maintaining
78 optimal body temperatures carry trade-offs (Huey & Slatkin, 1976). Direct costs, like
79 increased energy expenditure from active thermoregulation (Orrell et al., 2004), or indirect
80 costs, such as heightened predation risk (Herczeg et al., 2008; Skelly 1994), are likely to
81 increase as individuals seek micro-habitats to achieve optimal temperatures. According to the
82 cost-benefit model of thermoregulation (Huey & Slatkin, 1976), these trade-offs influence the
83 extent to which ectotherms thermoregulate. For example, thermoregulation should be more
84 precise when the environmental costs are low and the benefits of maintaining optimal body
85 temperatures are high (Vickers et al., 2011). This balance between the costs and benefits
86 shapes ectotherms' responses to their thermal environments (Sears et. al., 2016; Chan et al.
87 2024). Understanding these trade-offs is therefore crucial for predicting how animals optimise
88 body temperatures while minimising predation risks or additional environmental costs, and
89 how these adjustments correspond to fitness-related traits such as growth, reproduction, and
90 survival.

91 Variation in environmental conditions, particularly seasonal fluctuations, drive the
92 thermoregulatory decisions that ectotherms must navigate to obtain and maintain optimal
93 body temperatures in the wild (Giacometti et al., 2024). These fluctuations include not only
94 temperature changes but also shifts in water balance, food availability, predation pressures,
95 and interactions with conspecifics (Huey & Pianka, 1977; Leith et al., 2024). Seasonal shifts
96 change the physical and thermal landscape, altering the availability of suitable microhabitats

97 and thermal refuges in positive or negative ways (Sears & Angilletta, 2015). For
98 example, in high-cost environments where ectotherms must expend more time and
99 energy moving between microhabitats to optimise body temperature, individuals may
100 grow more slowly due to energy diverted to thermoregulation (Brewster et al., 2013)
101 or increase predation risk due to conspicuous behaviours to increase body temperature
102 (Basson et al., 2016). The behavioural responses to environmental fluctuations,
103 therefore, can help determine the physiological trade-offs that may influence survival
104 (Chan et al., 2024).

105 Heliothermic lizards thermoregulate through behavioural means by seeking or
106 avoiding heat sources, adjusting posture, and changing activity levels (Huey, 1982;
107 Porter et al., 1973). Key components of thermoregulation include the range of
108 individual thermal preferences available in the environment and the effectiveness of
109 maintaining body temperature through behavioural means. Physiological responses to
110 thermoregulation can be illustrated through a thermal performance curve (TPC),
111 which depicts the relationship between body temperature and performance (Angilletta,
112 2009; Taylor et al., 2021). Using this curve the thermal optimum and maximum
113 performance are identified, and are traits often indirectly related to individual survival
114 (Gilbert & Miles, 2017; Pearson & Warner, 2018). However, TPC relationships are
115 typically measured under controlled laboratory conditions, where stochastic
116 environmental factors (temperature fluctuations, predation pressures, and food
117 availability) are minimised or eliminated (Angilletta et al., 2002; Albuquerque et al.,
118 2023; Wild & Gienger, 2018). While such studies provide valuable insights, they often
119 fail to account for the complexities and variability experienced by ectotherms in
120 natural environments (Childress & Letcher, 2017). This disconnect creates a
121 significant knowledge gap, as it is unclear how laboratory-derived metrics translate to
122 true ecological relevance for individuals in the wild. It is usually challenging to
123 accurately measure individual survival in field settings due to the small size of these
124 animals or the rarity of capturing predation events. For example, survival estimations
125 are often limited to recapture events between sampling years (Gilbert & Miles, 2017;
126 Husak, 2006), which may not be an accurate assessment of individual survival over
127 shorter (i.e., seasonal) time frames. To gain a realistic understanding of the strategies
128 employed by ectotherms to maintain optimal body temperatures and the performance
129 benefits within their thermal performance curves, it is essential to conduct more field-
130 based research.

131 Here, we examine in a field setting the links between survival and common
132 metrics of thermal biology, specifically thermoregulatory behaviour and thermal
133 performance curves, in the Australian central bearded dragon (*Pogona vitticeps*). In
134 addition, we provide new insight into seasonal patterns of thermal ecology. Previous
135 laboratory work with *P. vitticeps* has shown their thermoregulatory behaviours in the
136 lab align with the cost-benefit model of thermoregulation (Cadena & Tattersall, 2009),
137 yet it is unknown how these changes in precision relate to their behaviour in nature.
138 We use temperature-sensitive radio transmitters equipped with accelerometers to
139 quantify activity and body temperature in the wild (Fig. 1), allowing us to generate
140 field-based thermal performance curves. In addition, by combining body temperatures
141 with recordings of the thermal landscape, we estimate seasonal thermoregulatory
142 behaviour (Fig. 1D). This approach enables us to examine how aspects of thermal
143 performance curves and thermoregulatory behaviours relate to individual survival
144 (Fig. 1E) during the reproductive season (spring) when this species experiences its

145 highest predation pressures (Wild et al., 2022). We predict that thermoregulation will vary
146 according to seasonal environmental costs (Huey & Slatkin, 1976). However, because of the
147 lack of studies that account for sex differences in thermoregulation (Christian & Weavers,
148 1996; Vickers et al., 2011), we predict limited differences in body temperatures between sex.
149 Finally, we predict individuals with high maximum performance will have a higher likelihood
150 of survival compared to those with low maximal performance (Christian & Tracy, 1981;
151 Gilbert & Miles, 2017). To date, no studies have continuously collected seasonal data on
152 individuals to investigate the impact of thermoregulation on performance and how these
153 trade-offs relate to survival in natural settings.

154 **2. Materials and methods**

155 *2.1 Preferred body temperature estimation (T_{set}) and body temperature calibration*

156 Preferred body temperature (T_{set}) trials were conducted on adult *P. vitticeps* that were either
157 captured from the study site or captive-bred descendants of wild-caught lizards from the study
158 region (see section 2.2 for region description). Trials were conducted in a temperature-
159 controlled (20 °C) room where internal body temperatures were measured using surgically-
160 implanted temperature loggers (Thermochron iButton, recording every 2min) as lizards
161 moved along a laboratory thermal gradient that ranged from 20°C to 40°C (Fig. 1A; see SI for
162 details on gradient and surgical protocols). Animals remained in the thermal gradient for a
163 minimum of 12h. Preferred body temperature was defined as the bounds of the central 50%
164 (i.e., the 25th and 75th quartiles) of body temperature in the thermal gradient (Hertz et al.,
165 1993). Linear models were used to determine differences in T_{set} bounds between sex.

166 To predict internal body temperature using external body temperatures (‘surface
167 temperatures’) in field settings, we examined the relationship between body temperature and
168 surface temperature in a subset of captive animals measured in the indoor thermal gradient
169 ($T_{b, Predict}$; Fig. 1A,B). This subset was equipped with a Pinpoint Beacon 250 (Lotek Ltd.,
170 Havelock North, NZ) that was placed in a custom-fit backpack harness (Wild et al., 2022).
171 Each unit housed a temperature data logger that recorded surface temperature every 2s, which
172 was averaged every 2min to pair with body temperature with iButton. Gradient methods
173 followed the same protocol described above. The relationship between body and surface
174 temperature was estimated using linear regression and paired t-test (surface vs. internal
175 temperature at each time point) to examine the degree to which surface temperature
176 underestimated or overestimated body temperature. The equation from the linear regression
177 between body and surface temperature was used for $T_{b, Predict}$ correction.

178 *2.2 Field study area and radiotelemetry*

179 Field work for this study was conducted in a 140km² nature reserve (Bowra Wildlife
180 Sanctuary) near Cunnamulla Queensland, Australia. Adult *P. vitticeps* were captured
181 opportunistically and tracked continuously between October 2018 to September 2019. Each
182 lizard was fitted with a Pinpoint Beacon 250 using the same custom-fit backpack harness used
183 in the $T_{b, Predict}$ experiment. Each unit housed a GPS logger, a single-stage VHF transmitter
184 (150–151 Hz), a temperature data logger, and a 2-axis accelerometer. Phenotypic sex was
185 determined using hemipenile eversion. For further information on lizard collection, site
186 description, or radio telemetry, see Wild et al. (2022).

187 *2.3 Field predicted body temperature, operative environmental temperature, and 188 thermoregulatory strategy*

189 Temperature dataloggers in the Pinpoint Beacon 250 measured the range of temperatures that
190 lizards experience in the wild. Loggers recorded a surface temperature (°C) every 2s, and this
191 was averaged over 1min. The surface and body temperature correction was applied (Fig. 1A)
192 to estimate field body temperature ($T_{b, Predict}$). Final body temperatures ($T_{b, Predict}$) for each

193 lizard were averaged for each hour between 0500-2100. Hourly core body temperatures
194 ($T_{b,obs}$) in the field were recorded for a subset of individuals (n=8) using surgically implanted
195 iButtons and showed a near one-to-one relationship between laboratory-adjusted ($T_{b,Predict}$)
196 and observed field temperatures ($p < 0.001$, *see* SI for results, Fig. S1).

197 Environmental temperatures available to animals within the landscape (T_e)
198 were estimated using physical models (Bakken & Gates, 1975) that were the same
199 length and width of an average lizard. Models were constructed of hollow copper
200 pipes (4.0cmx25.0cm) with a iButton suspended in the centre (Fig. 1C). These models
201 were validated by comparison with fresh lizard carcasses with implanted iButton
202 dataloggers recording internal body temperature (*see* SI for calibration methods).
203 Copper models were deployed from October 2018 to September 2019 and recorded
204 environmental temperature (T_e) every 1h. Copper models were placed in five micro-
205 habitat categories: full shade (n=10), partial shade (n=10), open (n=10), tree (n=12),
206 and burrow (n = 8; *see* Table S1 for definitions of micro-habitat categories). Micro-
207 habitats accessible to *P. vitticeps* were considered when positioning each model (*see*
208 SI for model calibration and verification). Mean T_e measurements were calculated for
209 each hour between 0500-2100 to obtain a measure of the operative temperature of the
210 habitat available to *P. vitticeps* for any given hour during the study. We assumed
211 males and females experienced the same distribution of thermal microhabitats. We
212 compared our T_e estimates to those predicted using the NicheMapR biophysical
213 modelling software (Kearney & Porter, 2017) and found a close and near one-to-one
214 relationship between our T_e estimates and those predicted with NicheMapR ($p < 0.001$,
215 *see* SI for results, Fig. S2).

216 Metrics of thermoregulation were quantified using laboratory preference range
217 (T_{set}) and hourly measurements of environmental (T_e) and body temperature ($T_{b,Predict}$)
218 in the field. The accuracy of thermoregulation (d_b) was defined as the overall mean
219 deviations of body temperatures from the thermal preference range (calculated using
220 sex-specific T_{set} values). Similarly, the average thermal quality of the habitat (d_e) was
221 assessed by estimating the overall mean deviations of environmental temperatures
222 from the thermal preference range for each individual copper model in each habitat
223 (Hertz et al., 1993). These metrics were calculated hourly between 0500-2100h across
224 the year. The hourly effectiveness of thermoregulation (E) for each individual lizard
225 was then calculated using d_b and d_e with the following equation:

$$226 \quad E = 1 - (\bar{d}_b) / \bar{d}_e$$

227

228 where E is expressed as a ratio generally ranging from 0 to 1, and over bars indicate mean
229 deviations of body and environmental temperature. An E of 1 reflects perfect
230 thermoregulation, where body temperature falls within the thermal preference range. An E of
231 0 is indicative of an animal not thermoregulating (Hertz et al., 1993). It is possible for E to be
232 negative in situations where an individual actively avoids the thermal preference range even
233 though T_e allows the opportunity for thermoregulation within the thermal preference range.
234 Low E values can occur when predators are abundant, food availability is scarce, or during
235 interaction with conspecifics (Christian & Weavers, 1996). All metrics of thermoregulation
236 ($T_{b,Predict}$, d_b , d_e , and E) were averaged for each individual over the course of each season prior
237 to analysis. For each metric, a linear mixed-effects model was used to test the effect of
238 season, sex, and their interactions, with season and sex as fixed effects and either lizard ID (or
239 model ID) as a random effect.

240 *2.4 Activity and thermal performance curves*

241 Activity (min/h) and field thermal performance curves (TPC) were estimated using the
242 accelerometry and temperature data provided by the Pinpoint Beacon 250. Accelerometers
243 recorded acceleration on two axes corresponding to X-heave and Y-surge at a rate of 6Hz.
244 Acceleration values were averaged for each axis (1min) between 0500-2100h for each season
245 (spring, summer, autumn, winter). Each axis of acceleration was transformed to resultant
246 acceleration (hereafter acceleration, ms^{-2}) following manufacturer protocols (*see* SI for
247 transformation details). Activity was defined as any change in acceleration from the previous
248 value between samples taken with the accelerometer and calculated as the minutes moved for
249 each hour (min/h). For analysis purposes activity was log-transformed ($\log(x+1)$) to deal with
250 the abundant sedentary periods in which individuals did not move (i.e., no changes in
251 acceleration).

252 Thermal performance curves were constructed using $T_{b,\text{Predict}}$ and acceleration (ms^{-2})
253 values from accelerometers. Body temperatures ($T_{b,\text{Predict}}$) were averaged for each 1min to
254 match the averaged timescale of acceleration data. General additive mixed-models (GAMM)
255 were used with $T_{b,\text{Predict}}$ as the predictor and acceleration (i.e., performance) as the response
256 variable. Performance for TPC was defined as the 95th percentile of acceleration at each 1°C.
257 This allowed the ability to characterise the upper capacity for movement while avoiding the
258 influence of outliers resulting from the many sedentary periods. The package *mgcv* was used
259 for cubic spline rolling average regression for all GAMM (Wood, 2017). Model selection,
260 fitting, and validation followed Zurr et al. (2009). The most inclusive GAMM included (in
261 addition to temperature) season, sex, and their interaction as fixed effects, and individual as a
262 random effect [modelled as a smoothed cubic spline]. The maximum predicted acceleration
263 (ms^{-2}) from GAMM fit was defined as P_{max} and the temperature associated with P_{max} was
264 defined as T_{opt} (Angilletta, 2009). For each TPC metric (P_{max} and T_{opt}) a linear mixed-effects
265 model was used to test the effect of season, sex, and their interactions, with season and sex as
266 fixed effects and lizard id random (repeated) effect. The *gam.check()* function from the
267 package *mgcv* was used to examine model convergence, gradient range, Hessian matrix
268 characteristics, and basis dimension checking results.

269 *2.5 Estimating survival*

270 Maximum likelihood survival probabilities were estimated using known fate models (White
271 & Burnham, 1999). Parameter estimates derived from known fate models were used to
272 determine the extent to which thermal or performance estimates could predict an individual's
273 survivorship in the field (Fig. 1E). Survival was determined from daily telemetry surveys
274 from Spring 2018, where deaths of individuals were documented when bodies were
275 recovered. Spring was used for this analysis because movement and mortality rates are
276 elevated during this period relative to other seasons (Wild et al., 2022), allowing the best
277 chance to link variation in thermal and performance estimates with survival outcomes. AICc
278 was used to correct for small sample sizes when estimating survivorship using known fate
279 models during the spring season, and models with ΔAICc of < 2.0 were considered to have
280 support. The analysis started with a fully saturated model in which survival probability during
281 the spring was dependent on movement (min/h), accuracy of thermoregulation (d_b),
282 effectiveness of thermoregulation (E), and maximum performance (P_{max}) as covariates, then a
283 series of reduced-parameter models were fitted where sex was included (or removed) as an
284 interaction.

285 *2.6 Statistical analysis*

286 Statistical analyses were performed using the R environment ver. 4.1.0 and survivorship
287 estimates using the program MARK (White & Burnham, 1999). All analyses were tested for
288 normality. If data did not fit normality assumptions, the appropriate transformation was

289 applied to achieve normality. Seasonal periods were spring, summer, autumn, and winter for
290 all analyses. Statistical significance was accepted at the $p < 0.05$, and if results were
291 significant, they were followed with the appropriate post-hoc test.

292 3. Results

293 3.1 Preferred body temperature estimation (T_{set}) and body temperature calibration

294 Preferred body temperature was determined for 20 lizards (male: $n=10$; female: $n=10$).

295 Females consistently had higher estimates than males. This was observed in the 75% quantile
296 measurements, with females at $33.8 \pm 0.92^\circ\text{C}$ and males at $29.0 \pm 0.92^\circ\text{C}$ ($F_{1,18}=4.78$; $p < 0.05$).

297 Similarly, in the 25% quantile measurements, females had estimates of $27.0 \pm 0.46^\circ\text{C}$, while
298 males had $25.5 \pm 0.46^\circ\text{C}$ ($F_{1,18}=4.77$; $p < 0.05$).

299 There was a strong relationship between laboratory body temperature and
300 surface temperature. (16,938 paired measurements were recorded for 10 individuals;
301 $R^2=0.94$; $F_{1,16937}=2,469,723$; $p < 0.01$). Surface temperature slightly overestimated body
302 temperature by $0.12 \pm 0.01^\circ\text{C}$ (paired $t=12.21$; $df=16,938$; $p < 0.01$), so body
303 temperature estimates ($T_{b,predict}$) were corrected from surface temperatures using the
304 linear regression results:

$$305 T_{b,Predict}=1.770+ (T_{surf} \cdot 1.058)$$

306 3.2 Thermoregulation in the field

307 Thermal-sensitive accelerometers were placed on 40 individual *P. vitticeps* (male: $n=32$;
308 female: $n=8$) that were tracked between Spring 2018 and Winter 2019. There were differences
309 in seasonal body temperatures ($T_{b,Predict}$) ($p < 0.01$) and differences between the season \times sex
310 interaction ($p < 0.01$; Table S2), but for sex alone there were no differences ($p=0.40$). Least
311 squares estimates indicated significant seasonal differences in $T_{b,Predict}$ (Table S3), with the
312 highest values in summer ($33.4 \pm 0.25^\circ\text{C}$), followed by spring ($29.2 \pm 0.27^\circ\text{C}$), autumn
313 ($26.5 \pm 0.25^\circ\text{C}$), and winter ($20.8 \pm 0.25^\circ\text{C}$). Least squares estimates for the interaction
314 suggested that differences in $T_{b,Predict}$ between the sexes were only observable during the
315 summer (Fig. 2A,B), where females selected higher body temperatures than males. There
316 were no detectable differences in $T_{b,Predict}$ during other seasons (Table S4).

317 Mean T_e was different across all seasons ($F_{3,329321}=371.03$; $p < 0.01$), with
318 higher temperatures observed in spring and summer, and lower temperatures in
319 autumn and winter (Fig. 2A,B). Season and the interaction between sex and season
320 had an effect on the accuracy of thermoregulation (d_b), but there was no overall effect
321 of sex on d_b estimates (Table S2). Males thermoregulated more accurately (i.e. low d_b)
322 than females during spring, and there were no differences during the other seasons
323 (Fig. 2C; $p < 0.05$). Season, sex, and the interaction had an overall effect on the thermal
324 quality of the habitat (d_e , Table S2). Thermal environment was more favourable (i.e.
325 lower d_e) for females than males during the summer (Fig. 2D; $p < 0.05$) because
326 females had a higher T_{pref} range than males.

327 The effectiveness of thermoregulation (E) was influenced by season, but the
328 effect of season was different between sexes (Table S2 & Fig. 3). In the spring season,
329 thermoconforming behaviours (i.e. low E) were exhibited by females, while males
330 displayed thermoregulating behaviours (i.e. high E). Both male and female lizards
331 engaged in thermoregulating behaviours during the summer (Fig. 3). However, in the
332 autumn and winter, lizards of both sex were less effective thermoregulators,
333 approaching thermoconformity to their environment (Fig. 3). Overall, the
334 effectiveness of thermoregulation (E) was higher (Table S2; $p=0.05$) in males (0.48)
335 than in females (0.29).

336 3.3 Seasonal activity and thermal performance curves

337 A total of 6,858,857 raw acceleration data points were collected on male (n=32) and
338 female (n=8) *P. vitticeps*. Average movement varied across the season ($F_{3,81}=9.25$; $p<0.01$),
339 but there were no differences between sexes ($F_{1,68}=0.23$; $p=0.63$) or the interaction
340 ($F_{3,81}=0.29$; $p=0.83$). Overall activity was highest in the summer and lowest in the winter (Fig.
341 4; Table S4).

342 The top candidate GAMM model for field thermal performance curves (ΔAIC score =
343 0.00) accounted for season, sex and their interaction allowing for random intercept and
344 smoothed spline per individual and explained 71% of the total deviance (Fig. 5; see S5 for
345 other model comparisons). Season ($F_{3,88}=190.62$; $p<0.01$) and the interaction between sex and
346 season ($F_{3,88}=143.08$; $p<0.01$) had an overall effect on the maximum performance, but there
347 was no effect on sex alone ($F_{1,90}=0.34$; $p=0.56$). Maximum movement performance (P_{max}) was
348 highest in spring, whereas winter had the lowest values of other seasons ($p<0.05$; Table S6).
349 Females exhibited higher P_{max} values in autumn and winter than in other seasons, and males
350 demonstrated higher values in spring and summer than in other seasons (Table S7). The
351 average thermal optimum (T_{opt}) temperature (mean \pm SE) was $36.6\pm 0.24^{\circ}\text{C}$. There were no
352 differences in T_{opt} across seasons ($F_{3,88}=0.24$; $p=0.63$), between sex ($F_{1,90}=0.57$; $p=0.64$), or
353 their interaction ($F_{3,88}=1.79$; $p=0.64$).

354 3.4 Applying metrics of thermoregulation, activity, and performance to survival

355 Twenty-seven lizards were tracked during the spring, eight of which died during this period.
356 Survival probabilities (mean \pm SE) were higher for males (0.75 ± 0.08) than females
357 (0.33 ± 0.20). The top competing model accounted for sex and maximum performance (Fig. 6;
358 Table S8). There was a distinct pattern between performance and survival for both sexes,
359 where individuals with lower maximum performance had higher survival rates compared to
360 those with higher performance. This decline happened at lower levels of P_{max} in females than
361 in males, such that a given P_{max} was associated with lower survival in females than males
362 (Fig. 6).

363 4. Discussion

364 In the context of the cost-benefit model of thermoregulation (Huey & Slatkin, 1976), our
365 study provides unprecedented insights into the trade-offs between thermoregulation and
366 survival in ectotherms. For the first time, we demonstrate that maximum movement
367 performance in the field is a strong predictor of individual survival, highlighting the nuanced
368 relationship between thermal biology and fitness outcomes. Using telemetry and temperature-
369 sensitive accelerometry, we generated the first in situ thermal performance curves for an
370 ectotherm, providing a rare examination of field-based thermoregulatory strategies and their
371 associated seasonal trade-offs. Notably, our findings reveal that high maximum performance,
372 often associated with optimal thermoregulation, correlates with an increased frequency of
373 predation for both males and females, with this effect being more pronounced in females. This
374 suggests a negative relationship between maximum movement performance and survival,
375 challenging the traditional view that higher performance within the thermal optimum
376 invariably enhances fitness. Our results underscore the critical role of individual behavioural
377 variation and sex in determining survival, and imply that seasonal shifts in thermoregulatory
378 behaviour and performance are key drivers of variation in field body temperatures,
379 thermoregulatory effectiveness, overall activity, and maximum performance observed in
380 natural settings (Angilletta et al., 2002; Husak & Fox, 2006; Sears et al., 2011).

381 Contrary to previous studies, which have linked maximum movement performance—
382 often measured as sprint speed under controlled laboratory conditions—to increased survival
383 (Gilbert & Miles, 2017; Pearson & Warner, 2018), we found that higher maximum
384 performance was associated with decreased survival in the wild. Our field-based findings

385 suggest that high-performing individuals may be more prone to engaging in conspicuous
386 or risky behaviours (Horváth et al., 2024), which could elevate their mortality rates due to
387 increased visibility, risk-taking, or age-related factors (Sergio et al., 2014; Stamps, 2007;
388 Ward-Fear et al., 2018). It is likely that individuals with greater acceleration that we
389 followed were engaging in more conspicuous activities, making them more susceptible to
390 predation (Skelly, 1994). Bolder lizards, which may exhibit high-performance traits
391 (Albuquerque et al., 2023), often have more extensive core home ranges and spend more
392 time in predator-abundant areas, increasing predation risk (Ward-Fear et al., 2018).

393 Outside of the reproductive season for females, we found that activity metrics of
394 thermoregulation and maximum performance followed general predictions of the cost-
395 benefit model of thermoregulation (Huey & Slatkin, 1976). Predictions from this model
396 postulate that ectotherms should prioritise thermoregulation when environmental costs
397 are low and the benefits of maintaining optimal body temperatures are high, while
398 reducing thermoregulatory efforts when costs outweigh the benefits. We observed that
399 during winter, when thermoregulation is more challenging due to lower ambient
400 temperatures and limited time to achieve thermal preference, there was a marked decline
401 in both the accuracy and effectiveness of thermoregulation. These declines coincided
402 with decreases in other physiological traits that are temperature-dependent, such as
403 maximum movement performance (ms^{-2}) and fine-scale activity (min/h). Conversely,
404 during the summer the accuracy and effectiveness of thermoregulation were high, which
405 corresponded with increased activity levels and maximum movement performance. These
406 seasonal trade-offs demonstrate the dynamic balance that lizards must maintain while
407 accounting for the energetic costs and benefits of thermoregulation (Angilletta & Sears,
408 2016; Vickers et al., 2011) and emphasise how seasonal thermoregulatory strategies
409 influence activity levels and performance, with direct implications for growth, survival,
410 and reproductive success.

411 Although the accuracy or effectiveness of thermoregulation was not critical for
412 survival in our study, such strategies of thermoregulation can play a significant role in
413 other aspects of lizard fitness, such as growth and maintenance (Brewster et al., 2013) or
414 preventing overheating (Huey & Slatkin, 1976; Kearney et al., 2009). Our results further
415 emphasise the seasonal dynamics of thermal biology, highlighting that maximising
416 physiological performance through active thermoregulation is a high priority. This is
417 particularly true for heliothermic lizards, which rely heavily on behavioural
418 thermoregulation and demonstrate the ability to adjust their thermoregulatory accuracy in
419 response to varying environmental costs (Sears et al., 2011).

420 Sex differences in thermal ecology in ectotherms have often been overlooked
421 (Bodensteiner et al., 2021; Huey & Pianka, 2007; Pottier et al., 2021) mostly because of
422 difficulties in studying the reproductive complexities of females. For example, females
423 often alter their behaviour and physiology during gestation, incurring additional costs not
424 experienced by males (Miles et al., 2000; Shine, 1980). For *P. vitticeps*, these trade-offs
425 likely contribute to the observed increases in daily movements and the reductions in body
426 condition during the reproductive season (Wild et al., 2022), as well as the overall higher
427 energy demands that females experience (Wild et al., 2023). These increased energetic
428 pressures associated with reproduction may drive the observed sex differences in the
429 effectiveness of thermoregulation during spring (Angilletta & Sears, 2000; Congdon,
430 1989). Although fine-scale activity did not differ between sexes, males and females may
431 allocate activity time differently (Orrell et al., 2004), allowing males to benefit more from
432 thermoregulation, while females focus on reproduction or predator evasion (Herczeg et

433 al., 2006; Gilbert et al., 2023). Differences in habitat use could explain differences in sex-
434 specific thermal efficiency due to varying microclimates (Logan et al., 2021). To our
435 knowledge, the present study is the first to account for sex when investigating the efficiency
436 of thermoregulation in natural settings for any ectotherm (*see* Herczeg et al., 2006 for lab
437 example). Comparing habitat use patterns or field metabolic rates between sexes could help
438 determine seasonal sex strategies of thermoregulation.

439 Laboratory results in other ectothermic vertebrates suggest a limited plasticity in optimal
440 temperatures (MacLean et al., 2019; Pottier et al., 2022; Zhang et al., 2023). Our field
441 findings corroborate this, indicating that the field optimal temperature of $36.6 \pm 0.24^\circ\text{C}$ —
442 where movement intensity was maximised—remained consistent across sexes and seasons. A
443 constrained thermal optimum suggests that energetically expensive behaviours, like
444 thermoregulation, are necessary to maintain optimal temperatures throughout the year,
445 regardless of environmental changes (Huey et al., 2012; Kearney et al., 2009). This
446 requirement becomes particularly challenging during energetically demanding periods like
447 spring and summer when heliothermic lizards divert surplus energy reserves toward
448 reproduction (Nagy, 1983). Maintaining a static thermal optimum appears to be crucial for
449 performance, despite the costs of thermoregulation (Herczeg et al., 2008; Huey & Slatkin,
450 1976). These findings highlight the trade-offs involved in maintaining optimal body
451 temperatures, as the energetic costs of thermoregulation must be balanced against other
452 physiological needs (Blouin-Demers & Nadeau, 2005; Vickers et al., 2011).

453 The methods used to quantify ectotherm responses to temperature gradients in the
454 laboratory are well-established (Angilletta, 2009; Huey, 1982; Taylor et al., 2021). However,
455 replicating laboratory methods in natural environments is challenging due to the extensive
456 data needed to link temperature fluctuations with behaviour and physiology, yet
457 understanding these interactions is crucial for insights into organism-temperature
458 relationships and selection processes (Angilletta, 2009; Christian & Tracy, 1981; Seebacher &
459 Franklin, 2005). For the first time in any lizard species, we have shown that temperature-
460 sensitive accelerometers can estimate thermal performance curves in a natural setting. We
461 refined the method of measuring body temperatures by calibrating body surface temperatures,
462 and our summer estimate ($32.7 \pm 0.02^\circ\text{C}$) aligns well with the only documented field core body
463 temperatures published for this species ($34.3 \pm 3.75^\circ\text{C}$: Greer, 1989; $32.9 \pm 0.88^\circ\text{C}$: Melville &
464 Schulte, 2001). New applications of accelerometers provide promising research avenues as
465 their usage continues to rise across other vertebrate groups (Chakravarty et al., 2019; Garde et
466 al., 2022; Pagano & Williams, 2019). Studies connecting data from accelerometers to
467 estimates of energy expenditure or identifying specific behaviours provide promising avenues
468 for future research regarding questions relating to thermal ecology.

469 We have shown that by combining basic physiological measurements with the
470 technology of thermosensitive accelerometers, it is possible to test challenging ecological and
471 physiological hypotheses regarding thermal ecology. Our approach allows the linking of
472 seasonal variation in thermal performance to survival, highlighting its role as a key driver.
473 Such analyses are crucial for predicting how species may respond to anthropogenic
474 disturbance and climate warming, and how selection may act on ectotherms in these scenarios
475 (Sinclair et al., 2016). Our data highlight the need to consider species-specific ecological
476 contexts when interpreting physiological metrics like sprint speed or acceleration, as selection
477 may act on ecological performance depending on how organisms utilise their maximal
478 performance in nature. Further work is needed to understand this dynamic relationship within
479 species and test how common physiological traits measured in the lab vary seasonally under
480 field conditions in a warming and more variable climate.

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492
493 **Conflict of interests**

494 We declare we have no competing interests.

495
496 **Author contributions**

497 K.W organised the sampling design, collection materials, laboratory work, and figures with
498 the support of J.R, J.C., P.P., L.S., A.G., and S.S. The first draft of the manuscript was written
499 by K.W. Comments from J.R, J.C., P.P., L.S., A.G., and S.S. contributed to the final version.
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502 L.S., Paul Waters and Jennifer Marshall Graves. The Ecological Society of Australia provided
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506 occurred.

507
508 **Data accessibility**

509 Data, code, and additional resources are available on GitHub: https://github.com/kris-wild/TPC_Survival.git

510
511
512
513 **Literature cited**

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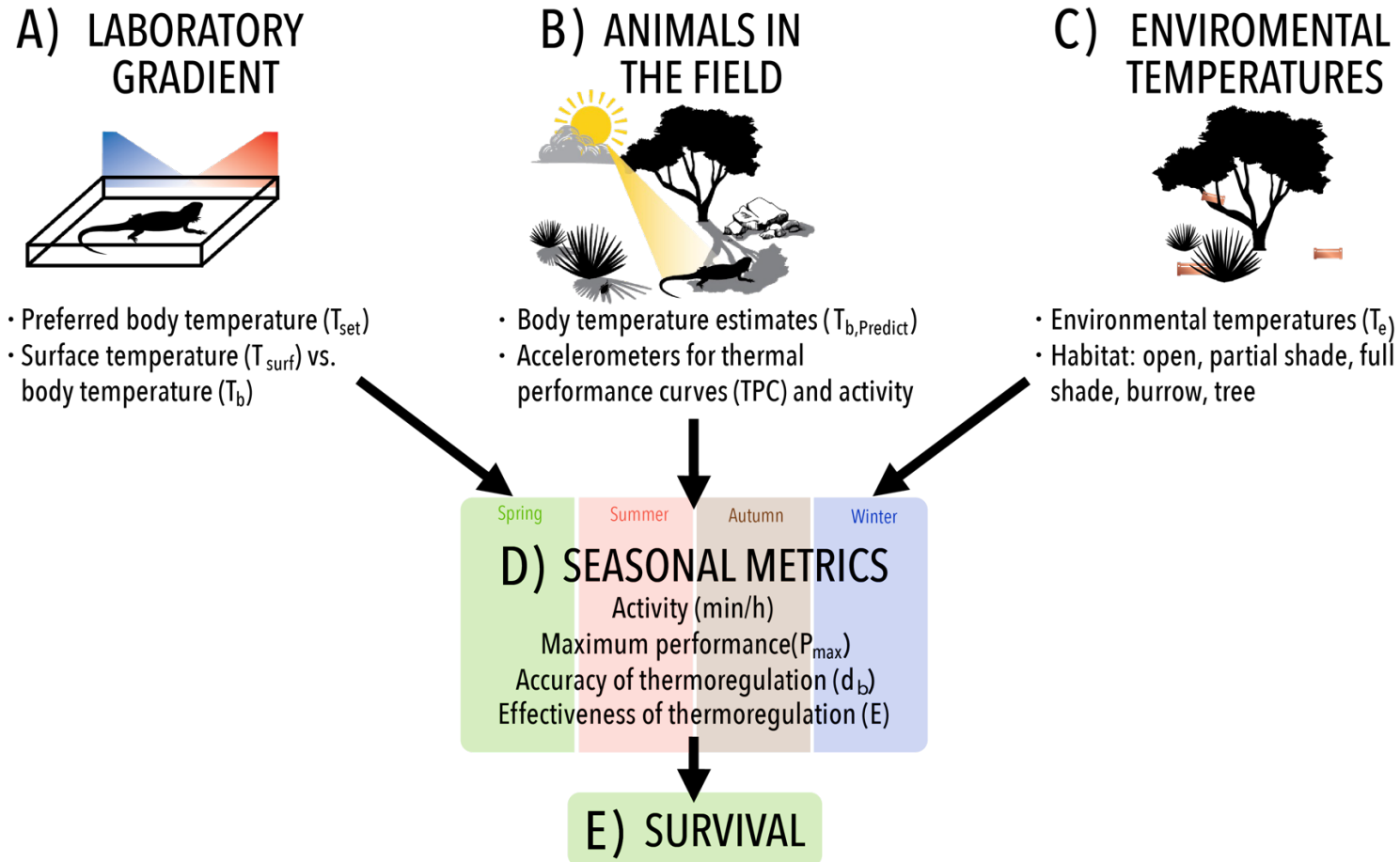


Figure 1. The comprehensive workflow of the experimental design aimed at identifying trade-offs in thermoregulation and their implications for survival. Laboratory thermal gradient experiments (A) were used to (1) measure the preferred body temperature (T_{set}) and (2) assess the relationship between surface temperatures (T_{surf}) recorded with accelerometers and internal body temperatures, enabling the prediction of body temperatures in the field ($T_{b, Predict}$). Seasonal thermoregulation and field performance metrics were evaluated using accelerometers (B). Copper pipes were placed in various microhabitats to characterise the thermal environment (T_e) available to lizards in the field (C). Metrics derived from experiments were then compared across seasons (D) and then used as covariates to understand their impact on survival during the spring season (E) when predation pressures are highest for this species.

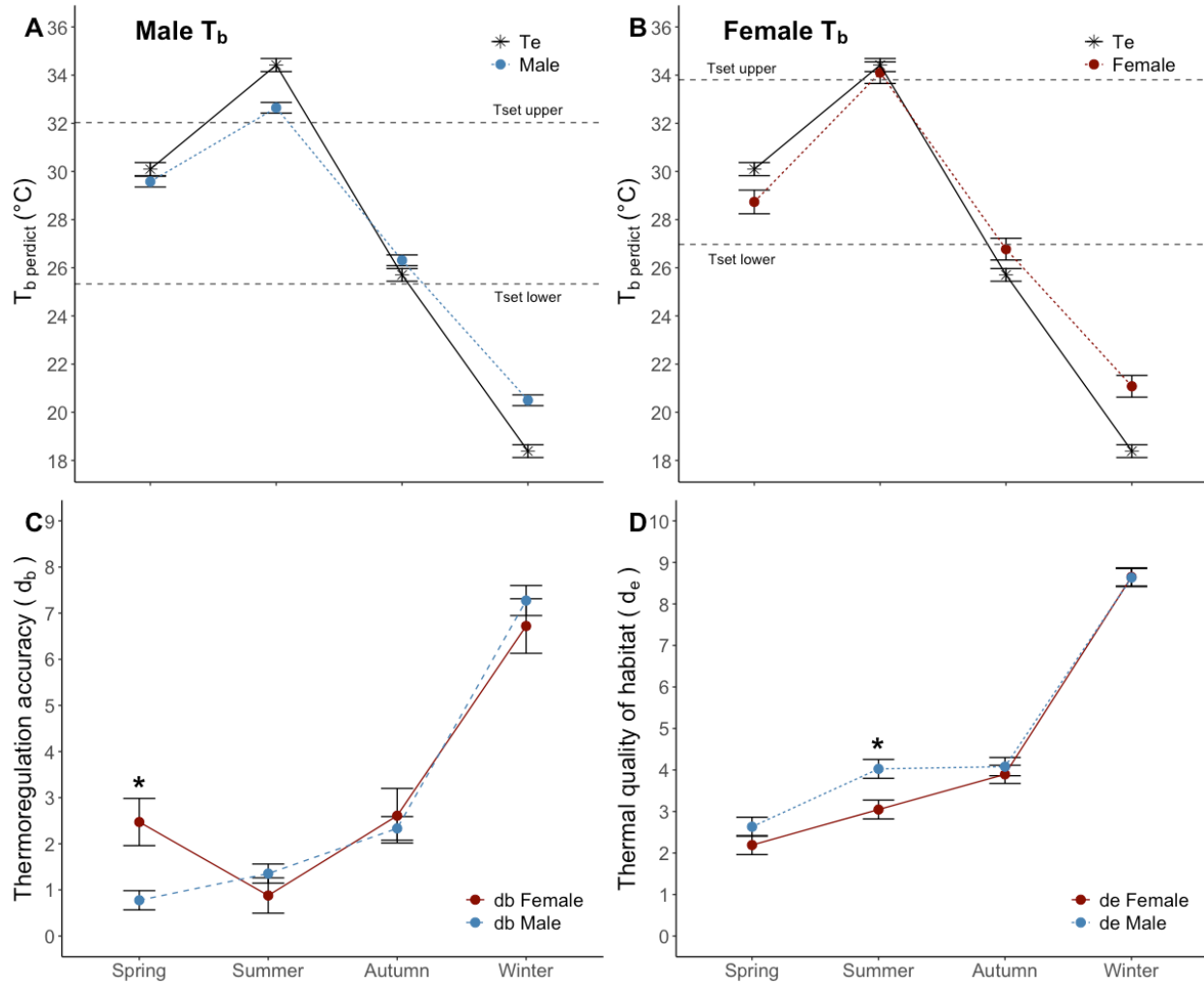


Figure 2. (A&B) Mean seasonal environmental temperature (T_e), thermal preference (T_{set}), and predicted body temperature ($T_{b,Predict}$) for male (A) and female (B) *Pogona vitticeps*. (C) Accuracy of thermoregulation (d_b) between sex, where low d_b denotes body temperature closer to thermal preference. (D) The thermal quality of habitat (d_e), measured with copper models, accounting for sex differences in thermal preference. Low d_e values indicate more environmental temperatures fell within T_{set} (i.e. favourable thermal environment). Error bars for all panels are ± 1 standard error of the mean. The asterisk symbol indicates a significant difference ($p < 0.01$) when comparing mean differences between sexes for that season.

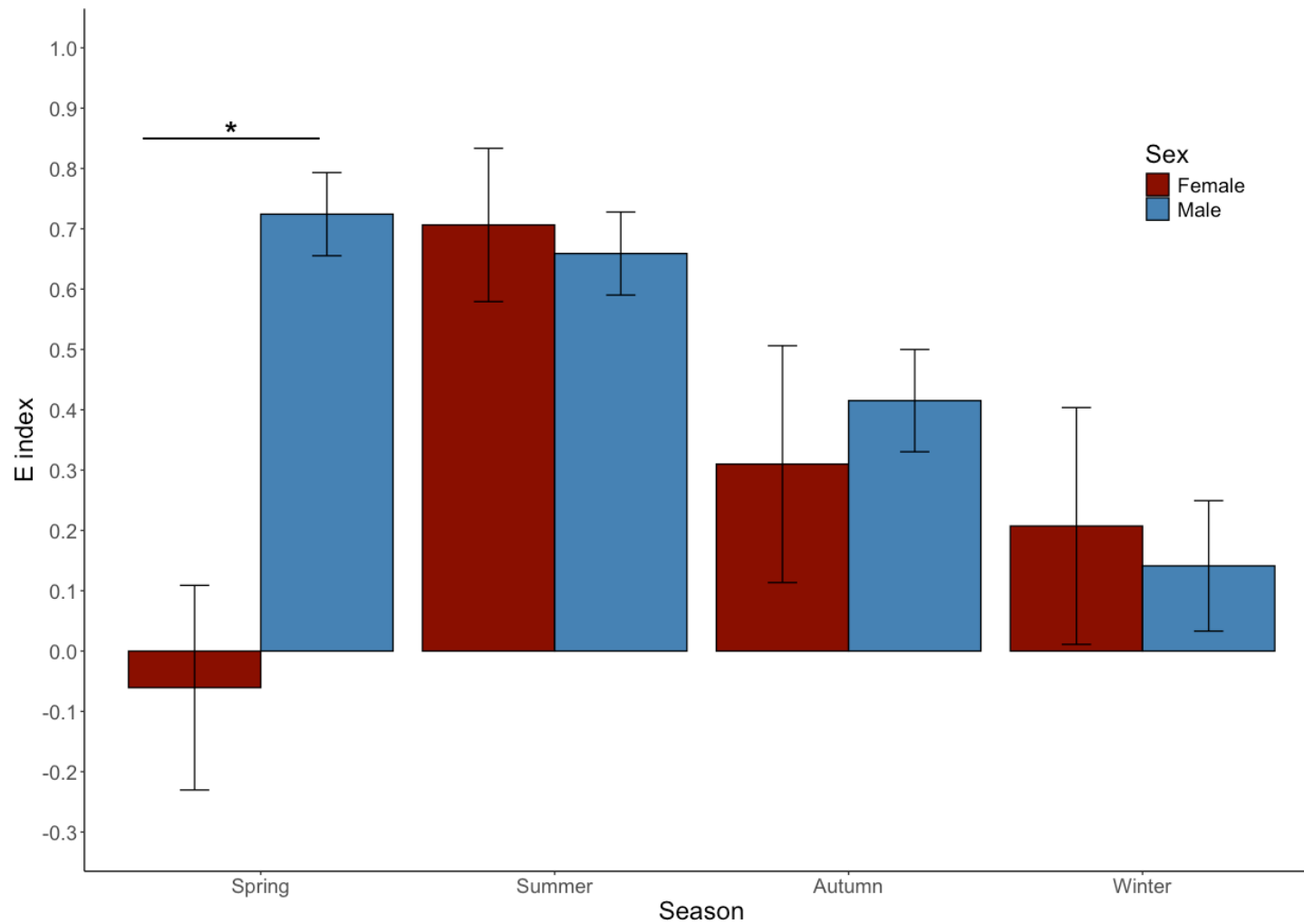


Figure 3. Effectiveness of thermoregulation (E index) by sex and season in *Pogona vitticeps*. E index values approaching 0 are indicative of conformity, and values approaching 1 are indicative of regulation. Values are means accounting for all individuals for each season. Error bars indicate ± 1 standard error of the mean. The asterisk symbol denotes a significant difference ($p < 0.01$) between sex when comparing mean differences for that season.

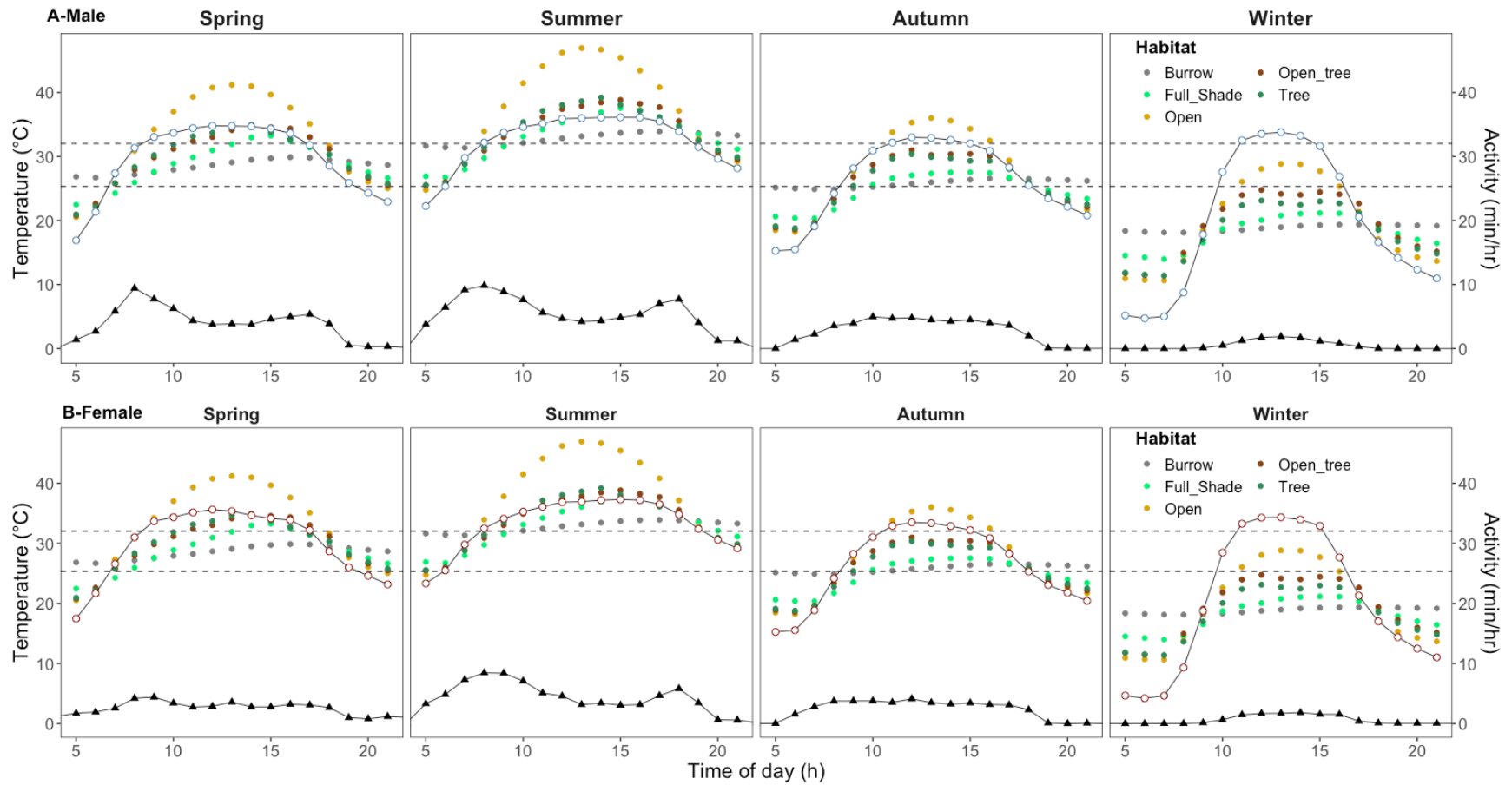


Figure 4. Mean predicted body temperatures (lines with circles) and activity levels (lines with triangles) for male (A) and female (B) *Pogona vitticeps* by season and time of day. The dashed line represents their preferred body temperature range. Coloured circles indicate mean environmental temperatures for different habitat types, measured using copper models.

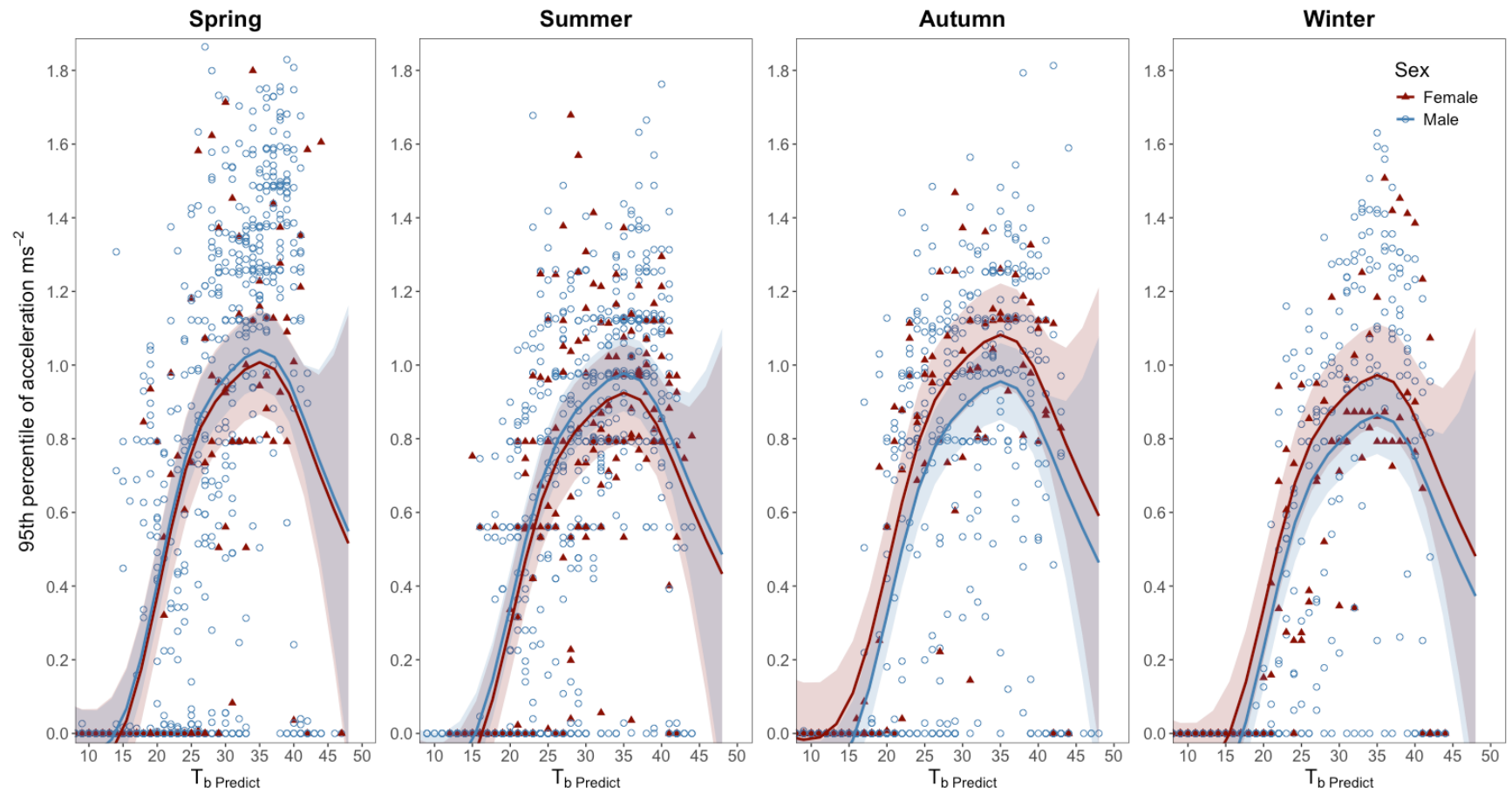


Figure 5. Thermal performance curves of free-ranging *Pogona vitticeps* across season and sex. The data were obtained from the top-performing Generalized Additive Mixed Models (GAMM) presented in Table S5. Each data point represents the average performance (95th percentile of acceleration) at a given temperature for all individuals in each season and sex. Bands around lines are 95%CI of model fit.

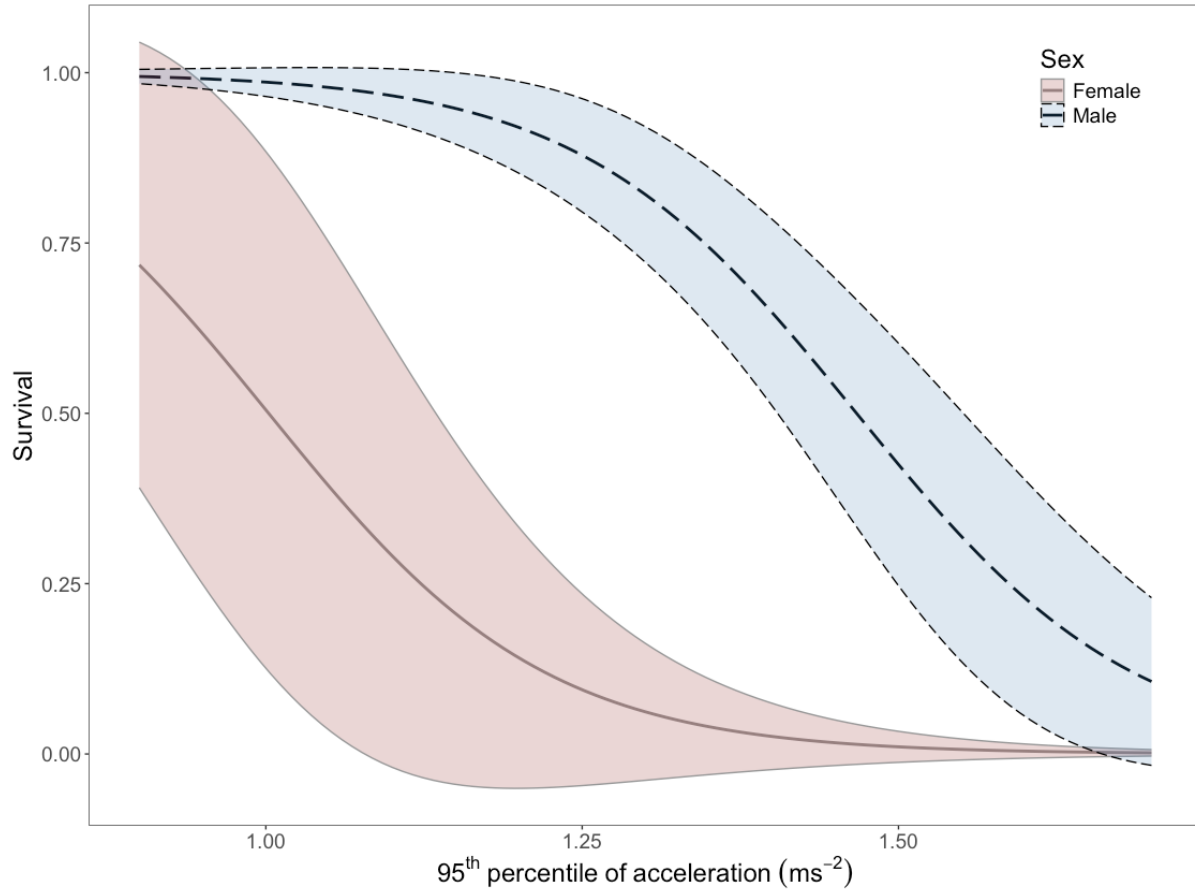


Figure 6. Survivorship as a function of the maximum performance (P_{\max}) for free-ranging male and female *Pogona vitticeps* in spring (September -November). Data are extracted from the top-performing survival model in Program MARK that accounted for season and sex (Table S8). Lines represent the predicted mean survival for each sex, and bands indicate 95%CI.

Supplement Information

Thermal gradient and surgical protocols:

Thermal gradient: A series of ceramic heat lamps placed above the gradient (5.0m long \times 1.0m high \times 2.0m wide) achieved continuous temperatures that ranged from 20°C to 40°C. The gradient was arranged in lanes where each sex was randomly assigned a lane before each trial. The thermal gradient contained sand (15cm depth) and fluorescent lighting that was on a 12h on/off cycle

Surgical protocols: Internal body temperature (T_b) was measured with a surgically implanted Thermochron iButton. To accurately measure internal body temperature (T_b), a Thermochron iButton was surgically implanted following the surgical techniques outlined by (Koenig et al., 2001). Each lizard (male: $n = 10$; female: $n = 10$) was given an inhalant anaesthetic (isoflurane 3 – 5) until the surgical plane of anaesthesia was reached. All iButtons were inserted into the peritoneal cavities by a 2cm incision through the ventral abdominal wall. Following surgery, lizards were allowed 48h to recover from surgery procedures prior to being placed in the gradient and then were placed in the gradient. The first 12h were considered an acclimation period once lizards were placed in the gradients before iButtons began recording T_b every 10min.

Body temperature validation in the field

Field body temperature vs laboratory body temperature adjustment: A subset of individuals in the field ($n = 8$) had iButtons that were surgically implanted following the surgical protocols described above. iButtons recorded hourly core body temperatures ($T_{b,obs}$) from January to March 2019. Accelerometers that recorded temperature were placed on these animals so that the laboratory $T_{b,predict}$ adjustment from surface temperatures could be compared to observed field core body temperatures. Comparisons of our body temperature adjustment ($T_{b,predict}$) and body temperature ($T_{b,obs}$) measured with surgically implanted iButtons revealed a close and near one-to-one relationship (regression statistics ± 1 se: slope = 0.82 ± 0.004 , intercept = 3.74 ± 0.140 , $r^2 = 0.86$, $N = 6,961$, $t = 203.87$, $p < 0.001$). It appeared that ($T_{b,predict}$) slightly under predicted core body temperature (Fig S1).

Operative model calibration and verification:

Operative model calibration: Models were calibrated using a fresh carcasses of *P. vitticeps*, which were placed beside one of the copper models on the ground in partial shade during three sunny days in November 2018. Temperatures were recorded in the carcass and the model every 5min from dawn to dusk. We used a linear regression of the carcass temperature to the model temperature to subsequently correct all records from field-deployed copper models (T_e).

Operative model verification: To further verify T_e values, we simulated operative temperatures of a non-thermoregulating animal in the open ($T_{e,predicted}$) using the package NicheMapR biophysical modelling software package (v.3.2.1) for the programming environment (4.2.2) to predict microclimates (Kearney & Porter, 2017). Historical weather data were obtained from the ERA5 reanalysis dataset (Klinges et al., 2022), a high-resolution ($0.1^\circ \times 0.1^\circ$, hourly) reanalysis product converted to microclimate model input using the pipelines described in (REF). This simulated biophysical model was set during the duration of our study (2018-2019) and was set at the geographic location of our study site (DD:145.679184, -28.094142). Specifically, the

'*micro_era5*' function was used to simulate microclimate, querying the soil grids database (soilgrids.org) for soil textural properties needed for soil moisture calculations, setting maximum shade to 80% and adjusting the local height for microclimatic estimates of air temperature, wind speed, and humidity to the body size of an adult *P. vitticeps*. The parameters of the model were based on the solar absorptivity documented in the field ($\alpha = 0.78$; Smith et al., 2016) and morphometric characteristics (mass:310g, SVL:25cm, and height:2.5cm) of an average adult *P. vitticeps*.

Comparisons of copper model T_e and $T_{e \text{ predicted}}$ ranged in temperature from -4.1° to 64.0°C revealed a close and near one-to-one relationship (Figure S1; regression statistics $\pm 1 \text{ se}$: slope = 1.052 ± 0.001 , intercept = 1.468 ± 0.105 , $r^2 = 0.95$, $N = 5533$, $t = 333$, $p < 0.001$). Therefore, we are confident that our T_e correction applies to various microhabitat types (Fig S2).

Accelerometer protocol and TPC analysis:

Calculation of Resultant Acceleration: To calculate the resultant acceleration, we considered only the x and y axes due to the limited acceleration on the z-axis for lizards. Resultant acceleration was computed using the Euclidean norm as follows:

$$\text{Resultant acceleration} = \sqrt{a_x^2 + a_y^2 + a_z^2}$$

where a_x , a_y and a_z are the accelerations along the x, y, z axes, respectively. The z-axis was ignored due to limited acceleration on that plane for lizards. This resultant acceleration provides a measure of the overall intensity of movement, integrating the contributions from both axes. This method ensures a comprehensive representation of the lizard's activity based on changes in acceleration.

Model selection TPC: Other GAMMs in the series considered all reduced variants of this model. This approach allowed us to compare Akaike Information Criteria (AIC) changes among models and allowed us to determine whether a given model explains significantly different amounts of the deviance in the data (Vickers et al., 2017). All GAMM models were ranked using AIC scores and those with ΔAIC of < 2.0 from the best model were considered to have support (Burnham & Anderson, 2004). The '*gam.check*' function was used to evaluate the adequacy of each model by examining model convergence, gradient range, Hessian matrix characteristics, and basis dimension checking results across multiple models. In general, TPC GAMM models showed a rise in the explanation of deviance when incorporating parameters that consider differences among individuals and season (Table S5).

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Supplementary figures & tables

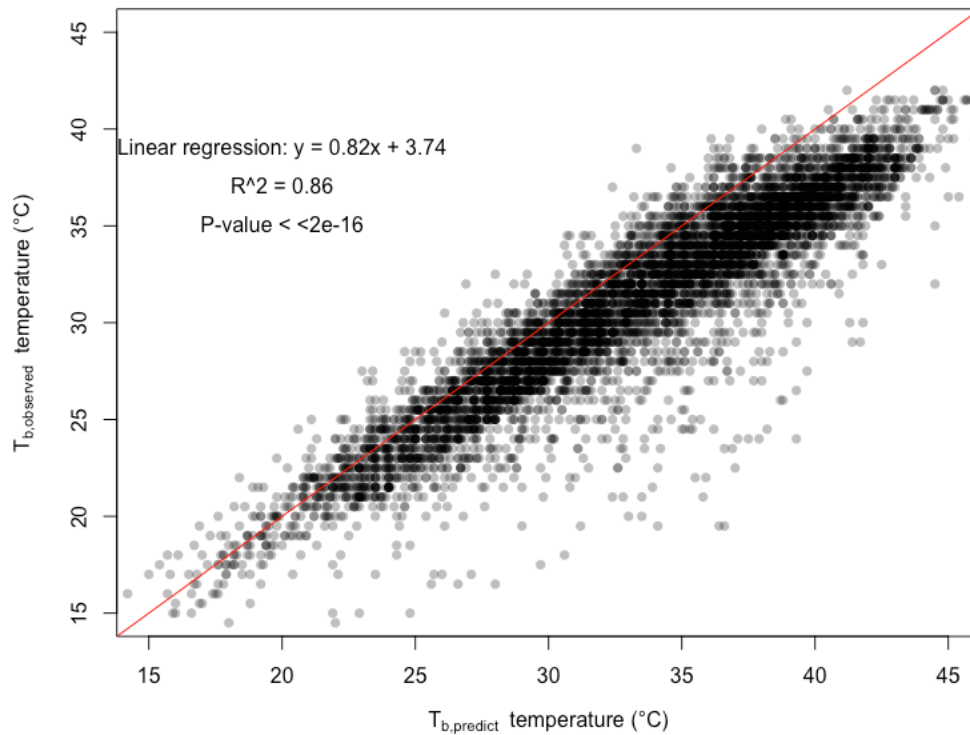


Figure S1. Comparison of predicted and core body temperatures of lizards in the field. Predicted body temperature ($T_{b,predict}$) was estimated through laboratory adjustments of surface temperature and core body temperature measured in a laboratory thermal gradient. Field core body temperature ($T_{b,observed}$) represents temperature recorded from implanted iButton in the field. The red line represents a perfect 1 to 1 relationship.

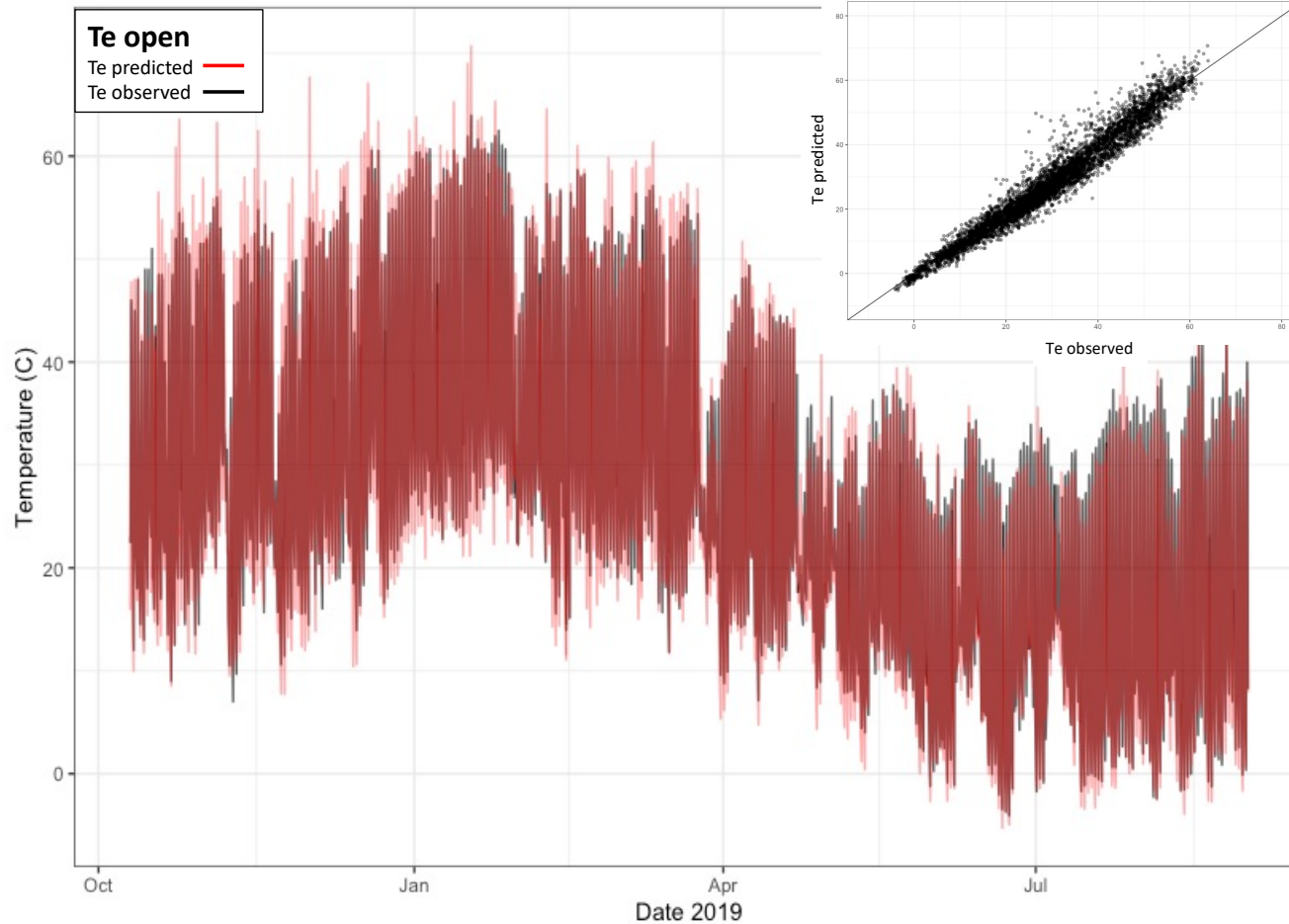


Figure S2. Hourly observations of lizard environmental temperatures measured with copper models T_e (black) and predicted T_e (red) across the entire study period from October 2018-September 2019. The panel in the top right indicate a 1:1 relationship between observed T_e and predicted T_e . Predicted T_e was estimated using the package NicheMapR (Kearney and Porter, 2017).

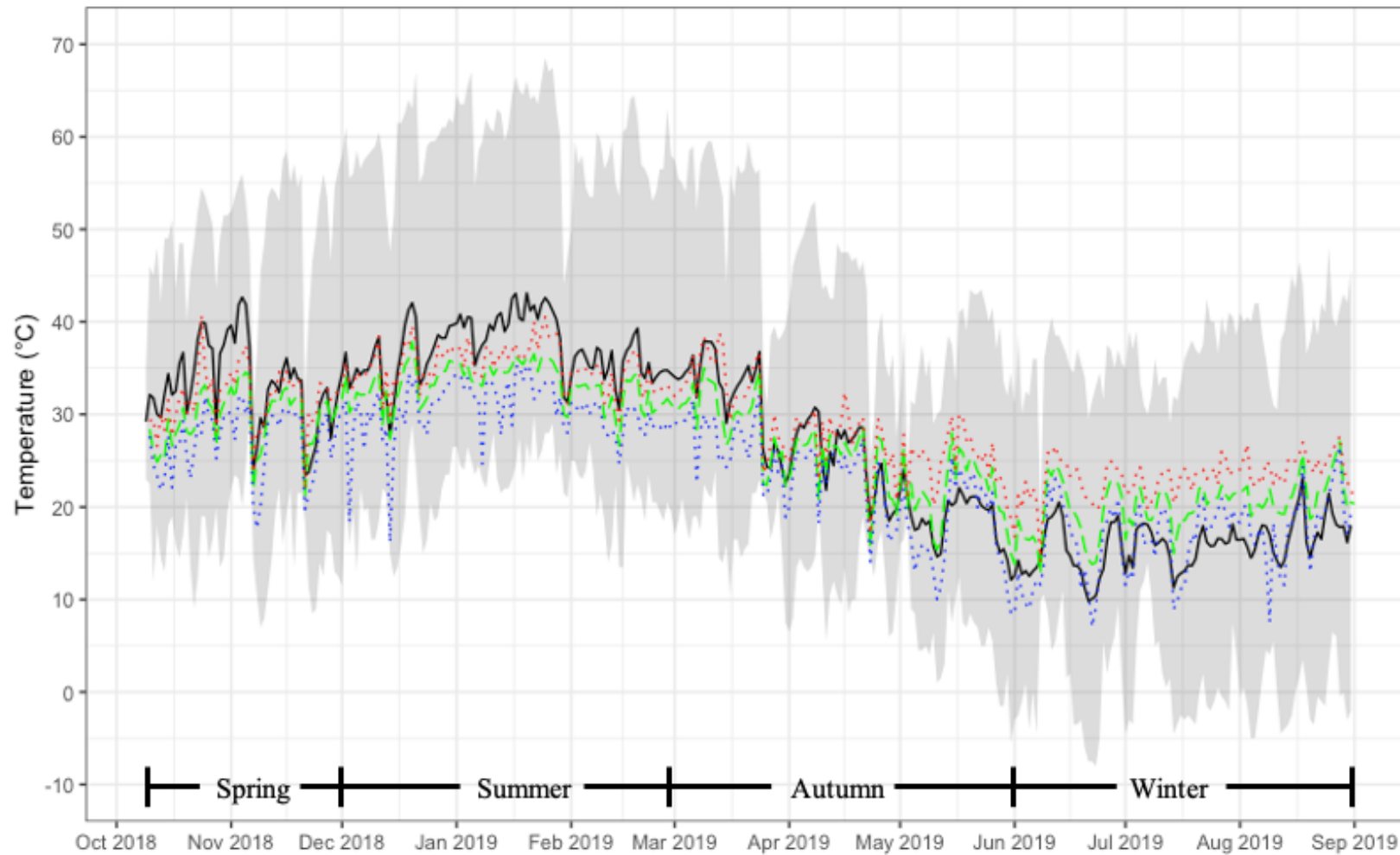


Figure S3. Environmental temperature range and how *Pogona vitticeps* thermoregulated during the duration of the study. Black solid lines represent the mean environmental temperatures of operative models (T_e) for each day, and grey bands represent the daily mean minimum and maximum of T_e . Coloured lines represent the daily mean (green), mean minimum (blue), and mean maximum (red) predicted body temperatures $T_{b \text{ Predict}}$ for all lizards during the study.

Table S1. Microhabitat categories of sun exposure. At each micro-habitat category, copper pipes were placed at each cardinal direction. Sun% was calculated using a spherical densiometer.

Exposure category	n	Definition
Full shade	10	%Sun < 25% on ground
Partial shade	10	25% ≥ % Sun ≤ 50% on ground
Full sun	10	%Sun > 50% on ground
Burrow	8	1m within open lizard/rabbit burrow
Shade-tree at 2m	4	Within shaded tree with %Sun > 50%
Partial shade-tree at 2m	4	Within tree 25% ≥ % Sun ≤ 50%
Open-tree at 2m	4	On branches of dead tree %Sun > 50%

Table S2. ANOVA table for predicted body temperature ($T_{b \text{ Predict}}$), accuracy of thermoregulation (d_b), thermal quality of habitat (d_e), and effectiveness of thermoregulation (E) for *Pogona vitticeps*. Each estimate is compared across the season, sex, and interaction. Individual lizard (or copper model ID) was treated as a repeated (random) variable. Bold values indicate significant differences.

Model Name	Effects	Sum Sq	Mean Sq	NumDF	DenDF	F value	p value
$T_{b \text{ Predict}}$	Sex	9.80	9.80	1	37	0.70	0.41
	Season	743,844.7 9	247,948.26	3	24,044	17,810.07	<0.01
	Season x Sex	1,814.79	604.93	3	24,044	43.45	<0.01
d_b	Sex	5.66	5.66	1	33	0.73	0.4
	Season	201,042.0 9	67,014.03	3	8,233	8,670.41	<0.01
	Season x Sex	816.86	272.29	3	8,233	35.23	<0.01
d_e	Sex	14.05	14.05	1	304.09	12.65	< 0.01
	Season	1983.80	661.27	3	306.74	595.59	< 0.01
	Season x Sex	12.36	4.12	3	304.09	3.71	0.01
E	Sex	0.46	0.46	1	83	4.10	0.05
	Season	2.14	0.71	3	83	6.34	<0.01
	Season x Sex	1.69	0.56	3	83	4.99	<0.01

Table S3. Tukey-Kramer multiple comparisons from $T_{b,predict}$ model (Table 2). Contrasts were extracted from the overall seasonal effect on $T_{b,predict}$.

contrast	estimate	SE	df	t.ratio	p value
Autumn – Spring	-2.61	0.15	10217.31	-18	<0.01
Autumn - Summer	-6.83	0.05	66587.12	-125.52	<0.01
Autumn - Winter	5.75	0.05	67846.58	125.52	<0.01
Spring - Summer	-4.22	0.14	9537.75	-29.2	<0.01
Spring - Winter	8.37	0.15	10266.14	57.53	<0.01
Summer - Winter	12.58	0.05	66582.53	229.03	<0.01

Table S4. Tukey-Kramer multiple comparisons of overall seasonal activity rate (min/h). Activity rate was $\log(x+1)$ transformed.

contrast	estimate	SE	df	t.ratio	p value
Spring - Summer	-0.28	0.15	92	-1.86	0.25
Spring - Autumn	0.12	0.20	83	0.61	0.93
Spring - Winter	0.80	0.21	83	3.89	<0.01
Summer - Autumn	0.40	0.20	82	2.03	0.19
Summer - Winter	1.08	0.21	82	5.23	<0.01
Autumn - Winter	0.68	0.23	69	3.01	0.02

Table S5. General additive mixed-models for investigating how performance curves varied across season, sex and their interactions for *Pogona vitticeps*. a) accounted for all individuals in the study, b) accounted for smooth per individual, c) accounted for sex as a fixed factor, d) accounted for sex as a fixed factor and allowed for smooth per individual, e) accounted for season as a fixed factor, f) accounted for season as a fixed factor and allowed for smooth per individual, g) accounted for season and sex as a fixed factor, h) accounted for season and sex as a fixed factor and allowed for smooth per individual, i) accounted for season, sex, and the interaction as a fixed factor, and j) accounted for season, sex, and the interaction as a fixed factor and allowed for smooth per individual. Models b:j accounted for random intercept for individual lizard. Bold values indicate values were considered to have support ($\Delta AICc$ of < 2.0).

Model id	Model	Residual Df	Residual Deviance	DF	AIC	Delta AIC	Deviance Explained (%)
j	Season + Sex + Season*Sex + s(Temperature, by = id) + (1 id)	2756.35	262.57	277.61	1688.84	0	70.57
h	Season + Sex + s(Temperature, by = id) + (1 id)	2760.13	264.88	273.83	1707.66	18.82	70.31
f	Season + s(Temperature, by = id) + (1 id)	2760.12	264.93	273.84	1707.87	19.03	70.3
e	Season s(Temperature) + (1 id)	2967.9	299.67	66.06	1724.75	35.91	66.41
d	Sex + s(Temperature, by = id) + (1 id)	2766.84	270.68	267.12	1760.11	71.27	69.66
b	s(Temperature) + (1 id)	2766.56	270.73	267.4	1760.93	72.09	69.65
i	Season + Sex + Season*Sex + s(Temperature) + (1 id)	2989.09	315.19	44.87	1840.57	151.73	64.67
g	Season + Sex + s(Temperature) + (1 id)	2992.13	317.28	41.84	1854.36	165.52	64.43
c	Sex + s(Temperature) + (1 id)	2986.68	319.93	47.28	1888.45	199.61	64.14
a	s(Temperature)	3033.96	358.89	8.96	2152.09	463.25	59.77

Table S6. Tukey-Kramer multiple comparisons from the Pmax model that accounted for the season, sex and interaction. Contrasts were extracted from the seasonal effect.

Contrast	Estimate	SE	df	t Ratio	p value
Autumn - Spring	-0.01	0.01	45	-1.3	0.57
Autumn - Summer	0.07	0.01	45	13.4	<0.01
Autumn - Winter	0.10	0.01	45	18.6	<0.01
Spring - Summer	0.08	0.00	45	15.1	<0.01
Spring - Winter	0.11	0.01	45	18.4	<0.01
Summer - Winter	0.03	0.01	45	6.1	<0.01

Table S7. Tukey-Kramer multiple comparisons from the Pmax model that accounted for the season, sex and interaction. Contrasts were extracted from season and sex interaction.

Contrast	Season	Estimate	SE	df	t Ratio	p value
Female - Male	Autumn	0.03	0.1	39	0.34	0.74
Female - Male	Spring	-0.13	0.1	38	-1.33	0.19
Female - Male	Summer	-0.15	0.1	38	-1.49	0.14
Female - Male	Winter	0.01	0.1	39	0.14	0.89

Table S8. Model comparisons of spring survival probability (ϕ) for *Pogona vitticeps*, depending on sex, movement (min/h), accuracy of thermoregulation (d_b), effectiveness of thermoregulation (E), and maximum performance (P_{max}). Sex interactions for d_b and E were accounted for because of the differences between males and females during the spring (Table S2). Values within the brackets are nested variables, and variables outside of brackets are covariates. Bold values indicate values were considered to have support ($\Delta AICc$ of < 2.0).

Model	AICc	$\Delta AICc$	AICc Weights	Model Likelihood	Number of Parameters	Deviance
$\phi(\text{Sex})P_{max}$	27.67	0.00	0.79	1.00	3	20.63
$\phi(.)P_{max}$	31.47	3.80	0.12	0.15	2	26.97
$\phi(.)$	34.99	7.30	0.02	0.03	1	32.82
$\phi(\text{Sex})$	35.31	7.64	0.02	0.01	2	30.81
$\phi(db)$	36.52	8.84	0.01	0.01	2	32.02
$\phi(.)E$	36.57	8.90	0.01	0.01	2	32.07
$\phi(.)T_{opt}$	36.86	9.18	0.01	0.01	2	32.36
$\phi(\text{Sex})d_b$	37.02	9.35	0.01	0.01	3	29.98
$\phi(.)\text{Activity}$	37.31	9.64	0.01	0.01	2	32.81
$\phi(\text{Sex})T_{opt}$	37.35	9.68	0.01	0.01	3	30.31
$\phi(\text{Sex})E$	37.36	9.68	0.01	0.01	3	30.31
$\phi(\text{Sex})\text{Activity}$	37.77	10.10	0.01	0.01	3	30.73