

1 **TITLE**

2 **Remote sensing restores predictability of ectotherm body temperature in the world's forests**

5 **RUNNING TITLE:**

6 Remote sensing and body temperature

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26 helpful guidance and feedback.

29 **BIOSKETCH**

30 **Adam C. Algar's** research asks how organisms respond ecologically and evolutionarily to climate
31 across spatial scales.

32 **Kate Morley's** MSc integrated remote sensing data into macrophysiology.

33 **Doreen S. Boyd's** main research interests are in the optimal exploitation of remote sensing systems to
34 understand ecosystem services provided by terrestrial systems.

41 **ABSTRACT**

42 **AIM:** Rising global temperatures are predicted to increase ectotherms' body temperatures, benefitting
43 some species but threatening others. Biophysical models predict a key role for shade in buffering
44 these effects, but the difficulty of measuring shade across broad spatial extents limits predictions of
45 ectotherms' thermal futures at the global scale. Here, we extend biophysical models of ectotherm
46 body temperature to include effects of forest canopy shade, via leaf area index, and test whether
47 considering remotely-sensed canopy density improves predictions of body temperature variation in
48 heavily shaded habitats.

49 **LOCATION:** Worldwide.

50 **TIME PERIOD:** 1990–2010.

51 **MAJOR TAXA STUDIED:** Lizards.

52 **METHODS:** We test predictions from biophysical ecological theory for how body temperature
53 should vary with microclimate for 269 lizard populations across open, semi-open, and closed habitats
54 worldwide. We extend existing biophysical models to incorporate canopy shade effects via leaf area
55 index, test whether body temperature varies with canopy density as predicted by theory, and evaluate
56 the extent to which incorporating canopy density improves model performance in heavily-shaded
57 areas.

58 **RESULTS:** We find that body temperatures in open habitats, like deserts, vary with air temperature
59 and incident solar radiation as predicted by biophysical equations, but these relationships break down
60 in forests, where body temperatures become unpredictable. Incorporating leaf area index into our
61 models revealed lower body temperatures in more heavily shaded environments, restoring the
62 predictability of body temperature in forests.

63 **CONCLUSIONS:** Although biophysical ecological theory can predict ectotherm body temperature in
64 open habitats, like deserts, these relationships decay in closed forests. Models incorporating remotely-
65 sensed data on canopy density improved predictability of body temperatures in these habitats,
66 providing an avenue to incorporate canopy shade effects into predictions of animals' vulnerability to
67 climate change. These results highlight the thermal threat of changes in canopy structure and loss of
68 forest cover for the world's ectotherms.

69

70 **KEYWORDS**

71 biophysical ecology, body temperature, canopy cover, land cover change, leaf area index, lizards,

72 macrophysiology, operative temperature, remote sensing, thermal ecology

73

74

75 **INTRODUCTION**

76 The implications of higher body temperatures in a warming world may be felt across all scales of life,
77 from metabolic rates (Dillon, Wang, Huey, 2010), to organismal behavior (Kearney, Shine, Porter,
78 2009; Sinervo, 2010), evolutionary fitness (Kingsolver, Diamond, Buckley, 2013), species'
79 distributions (Parmesan et al., 1999), and ecosystem dynamics (Cramer et al., 2001). Warm-adapted
80 species will benefit from hotter conditions, making a wider range of habitats available and
81 encouraging range expansion (Deutsch et al., 2008; Huey et al., 2009; but c.f. Logan, Huynh,
82 Precious, Calsbeek, 2013). In contrast, many species are already operating with slim thermal safety
83 margins, especially in biodiverse tropical environments (Deutsch et al., 2008; Huey et al., 2009;
84 Khaliq, Hof, Prinzing, Böhning-Gaese, Pfenninger, 2014; Sunday et al., 2014), suggesting that
85 future temperature increases will reduce activity times (Kearney et al., 2009), lower fitness
86 (Kingsolver et al., 2013), and increase the chance of extinction (Sinervo et al., 2010). These effects
87 may be especially severe in closed-canopy forests, where species tend to be cool-adapted (Deutsch et
88 al., 2008; Huey et al., 2009; Sunday, Bates, Dulvy, 2012). The effects of future warming on ectotherm
89 thermal vulnerability are often predicted using biophysical models of heat flux (Deutsch et al., 2008;
90 Kearney et al., 2009; Sinclair et al., 2016), but how well these models actually capture relationships
91 between microclimate and body temperature at the global scale over which predictions are made is
92 largely untested. Furthermore, these models do not currently incorporate the effects of forest canopy
93 shade on body temperatures, limiting their capacity to capture effects of global change on forest-
94 dwelling species.

95

96 Predicting the effects of climate warming and variability on organisms relies on understanding the
97 link between environmental conditions and body temperature (T_b). Although ectotherms must gain
98 their heat from the surrounding environment, standard climate variables, like mean annual air
99 temperature, are poor predictors of T_b worldwide (Meiri et al., 2013), not least because long-term
100 climate averages do not accurately reflect the microclimates experienced by individual organisms
101 (Kearney, Isaac, Porter, 2014; Kearney, Shamakhy, et al., 2014). T_b may also deviate from air
102 temperature, even when measured locally, for two reasons. Firstly, by behaviourally

103 thermoregulating, ectotherms can alter their T_b (Bogert, 1949; Huey, 1974; Huey & Slatkin, 1976).
104 Secondly, even for thermoconformers, T_b is not just a function of air temperature, but also depends on
105 the radiation absorbed and emitted by an animal, along with heat transfer via conduction and
106 convection and evaporative water loss (Porter & Gates, 1969; Gates, 1980; Bakken, Santee, Erskine,
107 1985; Campbell & Norman, 1998). The temperature that a non-thermoregulating animal would reach
108 at equilibrium in a particular environment is known as operative temperature (Bakken et al., 1985).
109
110 Shade plays a key role in determining operative temperature and species' thermal vulnerability to
111 climate change. By reducing the amount of solar radiation reaching an animal, shade can make the
112 difference between lethal and favorable body temperatures (Kearney et al., 2009; Sears et al., 2011;
113 Sunday et al., 2014) and alters the relative importance of different microclimate components: in full
114 sun, T_b will be sensitive to the intensity of incoming solar radiation, but in full shade, T_b should track
115 air temperature (Gates, 1980; Campbell & Norman, 1998; Buckley, 2008; Sears, Raskin, Angilletta,
116 2011). Shade will become increasingly important under future climate change, especially in tropical
117 regions, as animals must increasingly seek out cooler microhabitats to buffer against rising
118 temperatures (Kearney et al., 2009; Sunday et al., 2014). Despite its importance for the thermal future
119 of biodiversity, shade remains a significant challenge for predicting body temperature through space
120 and time. Currently, most models designed to predict operative and body temperatures of animals
121 predict a broad envelope of possible operative temperatures that encompasses full sun and full shade
122 or must assume a specific, spatially invariant, shade level (Kearney et al., 2009; Sunday et al., 2014;
123 Buckley, Ehrenberger, Angilletta, 2015). While shade from topographical features can be modelled
124 directly, provided detailed topographical information is available (Sears et al., 2011), quantitative
125 measures of shade from other sources—such as the forest canopy, which can greatly alter sub-canopy
126 thermal environments (George, Thompson, Faaborg, 2015; Frey et al., 2016; Lenoir, Hattab, Pierre,
127 2017)—is lacking from most models. Thus, current biophysical models, while effective in open
128 habitats, are likely to break down in forests, where shade is extensive (but not complete), limiting our
129 ability to accurately predict the thermal futures of species inhabiting these biodiverse environments.
130

131 In this paper, we extend existing biophysical models of ectotherm body temperature (Gates, 1980;
 132 Campbell & Norman, 1998; Buckley, 2008; Sears et al., 2011) to incorporate shade effects of the
 133 forest canopy, allowing for more precise predictions of ectotherm body temperature in forested
 134 environments using readily available remote sensing data on canopy density. Firstly, we model the
 135 expected relationships between body temperature, air temperature, solar radiation, and wind speed,
 136 ignoring canopy effects, and predict that model performance will decline across major habitat types
 137 with increasing shade levels, from barren lands to forests. Next, we use our extended model to
 138 evaluate whether incorporating remote sensing data to capture shade effects can improve model fit in
 139 heavily shaded forests.

140

141 **METHODS**

142 **Predicted relationships between body temperature and microclimate**

143 We generated predictions of how body temperature of a non-thermoregulating lizard will vary with
 144 microclimate (air temperature, incident solar radiation, and wind speed) in full sun by modelling
 145 operative temperature (T_e) using biophysical principles (Gates, 1980; Campbell & Norman, 1998),
 146 following Sears et al. (2011) and Buckley (2008):

$$147 \quad T_e = T_{air} + \frac{R_{solar} + R_{lw} - \varepsilon_s \sigma (T_{air} + 273)^4}{4\sigma (T_{air} + 273)^3 + c_p \left(1.4 + 0.135 \sqrt{\frac{v}{d}}\right)} \quad \text{Eq. 1}$$

148 where T_{air} is air temperature, R_{solar} is absorbed incoming solar radiation, R_{lw} is absorbed longwave
 149 radiation, ε_s is animal emissivity, σ is the Stefan-Boltzmann constant, c_p is the specific heat of air, d is
 150 the characteristic dimension of the animal and v is wind speed. The forest canopy will reduce the
 151 amount of solar radiation reaching an animal (Campbell & Norman, 1998), thereby lowering body
 152 temperature. We modelled the solar radiation incident on an animal as a function of the direct (beam)
 153 radiation, diffuse radiation, and the radiation reflected from the ground, following Buckley (2008) and
 154 Sears et al. (2011):

$$155 \quad R_{solar} = \alpha_s (F_p S_p + F_d S_d + F_r S_r) \quad \text{Eq. 2}$$

156 where α_s is the lizard's absorptivity of solar radiation, F_p , F_d , and F_g are view factors between the
 157 lizard and direct solar radiation (S_p), diffuse solar radiation (S_d), and reflected solar radiation (S_r),
 158 respectively. Direct and diffuse solar radiation are reduced under the forest canopy; we modelled
 159 direct solar radiation reaching an animal under the canopy ($S_{p,sub}$) using equations in Campbell and
 160 Norman (1998):

$$161 \quad S_{p,sub} = \omega_p S_p \quad \text{Eq 3}$$

162 where ω_p is the proportion of direct solar radiation that makes it through the canopy, which is an
 163 exponential function of LAI (Campbell & Norman, 1998):

$$164 \quad \omega_p = \exp(-\sqrt{\alpha_c} K_{b,z} LAI) \quad \text{Eq 4}$$

165 where α_c is the average absorptivity of the canopy and $K_{b,z}$ is the extinction coefficient for direct solar
 166 radiation at zenith angle z . Following Campbell and Norman (1998), we modelled the diffuse solar
 167 radiation under the canopy ($S_{d,sub}$) through numerical integration across all zenith angles (z):

$$168 \quad S_{d,sub} = S_d 2 \int_0^{\frac{\pi}{2}} \exp(-\sqrt{\alpha_c} K_{b,z} LAI) \sin z \cos z dz \quad \text{Eq 5}$$

169 The forest canopy also affects the amount of long-wave radiation reaching an animal from above
 170 (Webster et al., 2017). In the absence of a canopy, long-wave radiation from the air (L_a) is calculated
 171 as (Campbell & Norman, 1998; Buckley, 2008; Sears et al., 2011):

$$172 \quad L_a = \varepsilon_{ac} \sigma (T_{air} + 273)^4 \quad \text{Eq 6}$$

173 where ε_{ac} is clear-sky emissivity. Under a canopy, long-wave radiation ($L_{a,sub}$) comes from the sky and
 174 from the canopy, in proportion to the amount of clear sky (Webster, Rutter, Jonas, 2017):

$$175 \quad L_{a,sub} = V_s L_a + (1 - V_s) \varepsilon_c \sigma (T_{can} + 273)^4 \quad \text{Eq 7}$$

176 where T_{can} is canopy temperature, ε_c is canopy emissivity and V_s is a view factor denoting the
 177 proportion of long-wave radiation from clear sky. In Eq 4, the proportion of direct solar radiation
 178 reaching an animal through the canopy (ω_p) was modelled as a function of LAI (also see Essery,
 179 Pomeroy, Ellis, Link, 2008). As ω_p represents the proportion of radiation non-intercepted by the
 180 canopy, we derived the proportion of clear sky (V_s) using Equation 4 but assuming black leaves ($\alpha_c = 1$):

$$181 \quad V_s = \exp(-K_{b,z} LAI) \quad \text{Eq 8}$$

182 The forest canopy will also reflect longwave and reflected solar radiation from the substrate back
183 downward, with the process repeating, attenuated by the absorptivity of the canopy and the ground,
184 (Mahat & Tarboton, 2012), but we do not include this process here. We can now rewrite Eq 1 to
185 include the radiation reaching animals below the canopy as:

$$186 \quad T_{e,sub} = T_{air} + \frac{R_{solar,sub} + R_{lw,sub} - \varepsilon_s \sigma (T_{air} + 273)^4}{4\sigma (T_{air} + 273)^3 + c_p \left(1.4 + 0.135 \sqrt{\frac{v}{d}} \right)} \quad \text{Eq 9}$$

187 We used Eq 9 to model the predicted relationships between T_b and T_{air} , solar radiation, wind speed
188 and LAI. Full model details and parameter values are given in Appendix S1 in Supporting
189 Information. We varied each environmental variable across the range of values observed in our
190 empirical data to generate predictions of the shape of the relationships between each variable and T_b .
191 We modelled the relationships between T_b and T_{air} , solar radiation, and wind speed at five different
192 LAI levels to visualize interactive effects between these variables. We stress that our aim was not to
193 predict the absolute values of T_b in our empirical dataset, but rather the relationships that emerge
194 between T_b , microclimate, and LAI.

195

196 **Empirical data**

197 We tested whether global relationships between ectotherm body temperature, microclimate and LAI
198 matched those predicted by our biophysical model using a dataset, collected from the literature, of
199 mean body temperatures for 269 diurnal, non-fossorial mainland lizard populations (179 species; Fig.
200 S1.1) sampled between 1990 and 2010, building on an existing database (Clusella-Trullas, Blackburn,
201 Chown, 2011; Meiri et al., 2013). A list of data sources is found in Appendix 1. Following Meiri et al.
202 (2013), we did not set a minimum sample size, pooled data across sexes and life stages, and excluded
203 temperatures or populations sampled at night. We limited our data to post-1990 to limit confounding
204 effects of substantial 20th century land cover change on our estimates of canopy structure (see below).
205 For each population, we extracted the mean daytime (06:00–18:00 local time) air temperature 1cm
206 above ground (averaged across rock, soil, and sand), air temperature at 1.2m, solar radiation, and
207 wind speed at 1cm from the microclim dataset (Kearney, Isaac, et al., 2014) for the sampling months

208 reported for each study. We extracted microclimate data values in full sun for open and semi-open
209 habitats and in full shade for closed forests. We assigned each population to one of these habitat types
210 based on the geographical coordinates reported for a population. While this may not capture
211 microhabitat preferences perfectly (e.g. forest gap or edge specialists), this is likely to add noise to our
212 analysis rather than bias it. We made these assignments (Table S1.2) based on combining categories
213 included in the global land cover consensus product (Tuanmu & Jetz, 2014) and finding the land
214 cover (closed, semi-open, open) with the highest probability of occurring at each location, resulting in
215 27 closed, 123 semi-open, and 119 open habitat populations (Fig. S1.1).

216

217 We determined whether species were ground-dwelling, arboreal or semi-arboreal (use ground and
218 trees, or use shrubs) from the literature, using the source paper for the T_b data where possible and
219 other literature or expert knowledge where necessary. Data sources are given in Appendix 1 and the
220 accompanying dataset. For ground-dwellers, we used air temperature and wind speed at 1cm for all
221 analyses, for arboreal species we used air temperature and wind speed at 1.2m and for semi-arboreal
222 species, we used the average of 1cm and 1.2m. We inferred wind speed at 1.2m using the equations
223 given for this purpose in Kearney, Isaac, et al. (2014). We did not consider variation in substrate
224 temperature because in our empirical dataset (see below), air and soil temperature were highly
225 correlated ($r=0.94$, $P \ll 0.001$). If sampling months were not reported we used a summer average
226 (northern hemisphere: April–September, southern hemisphere: October–March) instead. For
227 populations whose coordinates fell in the ocean (likely due to georeferencing error), we used
228 conditions from the nearest piece of land, provided it was within 1 grid cell (at 0.1667×0.1667 DD
229 resolution) of the original coordinates.

230

231 We also extracted leaf area index (LAI) for each population's location. We used the mean of 8-day
232 MODIS reprocessed composites of LAI (Yuan, Dai, Xiao, Ji, Shangguan, 2011) across the sampling
233 months at 30 arc-sec spatial resolution, averaged from 2001-2010. LAI data are freely available from
234 <http://globalchange.bnu.edu.cn/research/lai/>. There is a partial temporal mismatch between our T_b data

235 (1990-2010) and our LAI data (2001-2010) that represents a tradeoff between limiting effects of land
236 cover change on LAI estimates for lizard sampling locations and maintaining sample size.

237

238 **Testing for predicted relationships**

239 We initially fit regression models that excluded LAI and modelled T_b as linear functions of air
240 temperature and solar radiation and quadratic functions of wind speed to approximate the predicted
241 non-linearity in the T_b wind speed relationships (Eqs 1 and 9). Wind speed data were right-skewed, so
242 we fifth-root transformed this variable to reduce high leverage of large values. We did not use a log-
243 transformation to avoid taking the logarithm of zero. We removed non-significant terms from our
244 model sequentially, based on P-value, starting with quadratic terms. Next, we added LAI to our full
245 regression model, including a quadratic term to capture non-linearity, as well as interactions with
246 wind speed and solar radiation. We removed non-significant terms as above, beginning with
247 interactions, then quadratic terms. LAI was also right-skewed so we square-root transformed it to
248 reduce influence of a few large values. We repeated regressions for all data habitats pooled and each
249 habitat (open, semi-open, closed) separately.

250

251 We fit all regressions using the `lmekin` function in the `coxme` package (Therneau, 2015) in R 3.4.3 (R
252 Core Team, 2018) to simultaneously incorporate spatial and phylogenetic non-independence into our
253 regressions (Freckleton & Jetz, 2009). Based on Tonini, Beard, Ferreira, Jetz, Pyron's (2016)
254 consensus tree, we used shared branch length between species (or populations, see below) as our
255 measure of phylogenetic covariance and the inverse of distance between locations as our spatial
256 covariance matrix. We also considered an alternative phylogenetic covariance structure by
257 transforming the tree using Pagel's λ of T_b and then recalculating shared branch lengths. Where
258 multiple populations of the same species from different locations were included in our data, we
259 replaced a species' terminal branch with a randomly resolved clade whose crown node depth was
260 chosen from a random uniform distribution with a maximum length equal to the original terminal
261 branch. Populations were added to the tree prior to pruning species not included in our analysis and
262 thus are represented as branching events which occurred after a species diverged from its sister. After

263 fitting regressions, we calculated the marginal (R^2_m) and conditional (R^2_c) R^2 values (Nakagawa &
264 Schielzeth, 2013). R^2_m is the proportion of variance explained by environmental variables and R^2_c is
265 the variance explained by environment, space and phylogeny.

266

267 **Geographical bias**

268 Our T_b data were highly geographically biased toward South America (Fig. S1.1). To determine how
269 this may have affected our results, we randomly sub-sampled the South American data and refit the
270 final model for each habitat type and all habitat types combined. For each regression, we reduced the
271 number of South American data points so that they were equal in number to the next highest continent
272 (Table 1). We repeated this process 1000 times and computed the number of times we detected a
273 significant ($P < 0.05$) relationship in the direction matching that found in the original regression. We
274 did not geographically subsample closed habitats because only three populations were not from South
275 America.

276

277 **RESULTS**

278 **Biophysical predictions**

279 The model based on environmental biophysical principles for a non-thermoregulating lizard in full
280 sun ($LAI=0$); Eq. 9) predicts that T_b will increase linearly with air temperature and solar radiation, and
281 that it will decline proportionally to the inverse of the square root of wind speed (Fig. 1). T_b is
282 predicted to decline in a nearly exponential fashion with increasing LAI (Fig 1) and to have
283 interactive effects on T_b with wind speed and solar radiation, but not air temperature (Fig 2). As LAI
284 increases, solar radiation above the canopy and wind speed both are predicted to have diminishing
285 influence on T_b (Fig. 2).

286

287 **Microclimate– T_b relationships**

288 As predicted, and after accounting for phylogenetic and spatial autocorrelation, we found positive,
289 global relationships between T_b and air temperature (slope \pm s.e. = 0.21 ± 0.03 , $P = 6.5 \times 10^{-10}$, Table S1.3)

290 and T_b and solar radiation (slope \pm s.e.=0.006 \pm 0.003, $P=0.02$, Table S1.3). Contrary to our prediction,
291 we found no relationship between T_b and wind speed (linear slope \pm s.e.=-2.6 \pm 2.4, $P=0.27$, Table
292 S1.3). Although relationships with air temperature and solar radiation were significant, these variables
293 explained just 13% of the variance in T_b worldwide (Table S1.3). Regression results using a Pagel's λ
294 transformed tree were nearly identical with no changes in direction of coefficients, R^2_m , or
295 significance (Table S1.4).

296

297 As predicted by biophysical models, T_b in open habitats, where there is less extensive shade, was
298 significantly related air temperature and solar radiation ($P<0.005$ for both; Fig. 3; Table S1.5). We
299 found no relationship with wind speed ($P=0.11$; Fig. 3; Table S1.5). In total, temperature and solar
300 radiation explained 35% of the variance in T_b (Table S1.5). The variance explained by microclimate
301 declined to 10% in semi-open habitats (Table S1.6), where there was no relationship between T_b and
302 solar radiation or wind speed ($P>0.7$ for both; Table S1.6), but the significant relationship with air
303 temperature was retained (slope \pm s.e.=0.18 \pm 0.05, $P=0.0003$, Fig. 3). These relationships did not
304 change when a Pagel's λ -transformed phylogeny was used (Table S.4)

305

306 In shade-dominated closed forests, we found no significant relationships between T_b and air
307 temperature, solar radiation, or wind speed (Fig. 3; Table S1.7; $P>0.25$ in all cases). The low sample
308 size ($n=27$) in closed forests compared to semi-open ($n=123$) and open habitats ($n=119$) results in low
309 statistical power. In closed forests, to achieve significance at $P<0.05$ with statistical power of 0.8, we
310 would have needed sample sizes of 569 (air temperature), 362 (solar radiation), or 193 (wind speed),
311 suggesting that the lack of relationships in closed forests is not simply a function of lower power
312 compared to open and semi-open habitats. The variance in T_b explained by microclimate including air
313 temperature, solar radiation and a linear wind speed term was only 8% (Table S1.7). These results
314 were insensitive to using a Pagel's λ -transformed phylogeny (Table S4).

315

316 **LAI and T_b relationships**

317 We found no global relationship between T_b and LAI, either as a main effect or as interactions with
318 solar radiation and wind speed when all land cover types were pooled ($P > 0.25$ in all cases; Fig. 4;
319 Table S1.3). However, as predicted, T_b was negatively related to LAI in closed habitats (Fig. 4; Table
320 S1.3; slope \pm s.e. = $-5.0 \pm 0.2.3$, $P = 0.03$). We found no significant interactions between LAI and wind
321 speed or solar radiation in these habitats (Table S1.3). Including LAI in regressions of T_b on micro-
322 environmental conditions in closed forests almost tripled the variance explained by microclimate (T_{air} ,
323 solar radiation and linear wind speed) alone from 7% to 19% (Fig. 4; Table S1.7). In contrast, LAI
324 explained very little additional variance in T_b in semi-open and open habitats or when all data were
325 pooled ($< 2\%$; Fig. 4; Tables S1.3, S1.6, S1.7), signifying it is of little to no importance for T_b in these
326 habitats. None of these regression results were affected by using a Pagel's λ -transformed phylogeny
327 (Table S1.4).

328

329 **Geographical bias.**

330 Subsampling our South America data revealed that relationships between T_b and air temperature were
331 robust to geographical bias in our dataset (Table 1). Relationships with solar radiation were variable
332 when all habitat types were pooled, but their direction was consistent in open habitats, though only
333 significant 11% of the time (Table 1).

334

335 **DISCUSSION**

336 We identified a systematic decay in the ability of existing biophysical models to predict global
337 variation in body temperature (T_b) across major habitat types as a function of shade availability, from
338 open habitats where models performed well to closed forests where T_b was unpredictable.

339 Relationships between body temperature, air temperature and solar radiation matched predictions
340 from classic biophysical models in open habitats across the globe, but the same models failed in
341 forests because even estimates of air temperature from full shade fail to accurately capture thermal
342 conditions under the canopy. By extending biophysical models of ectotherm heat flux to incorporate
343 the interception of solar radiation by the forest canopy, we showed it is possible to predict ectotherm
344 body temperature variation in forests using readily available remote sensing data. As predicted by our

345 model, we found that in closed forests, body temperatures cooled with increasing canopy density.
346 Moreover, including LAI in T_b -microclimate regressions almost tripled the variance explained by in
347 closed forests, revealing potential to improve predictions of body temperatures of forest species under
348 future climate change and highlighting the importance of future canopy thinning and loss for the
349 thermal future of forest ectotherms.

350

351 The relative importance of different microclimate variables for T_b varied across major habitat types.
352 In open habitats, T_b was sensitive to air temperature and solar radiation, as predicted by biophysical
353 models (Gates, 1980; Campbell & Norman, 1998; Buckley, 2008; Sears et al., 2011). These
354 relationships reflect proximate effects of microclimate on heating and cooling of organisms where
355 shade is too rare or unevenly distributed to permit efficient thermoregulation to lower temperatures
356 (Huey, 1974; Huey & Slatkin, 1976; Sears et al., 2016), but also capture longer-term adaptive
357 responses that have resulted in higher preferred temperatures in open environments, especially if these
358 environments are also drier (Clusella-Trullas et al., 2011). Contrary to biophysical predictions, we
359 found no relationship with wind speed in open habitats which could either reflect lower accuracy of
360 wind data, or perhaps reduced activity of lizards in windier conditions (e.g. Logan, Fernandez,
361 Calsbeek, 2015). The sensitivity of the solar radiation (but not air temperature) relationship to sub-
362 sampling of the South American data indicate that it has a weaker effect than temperature and thus its
363 detectability may be more prone to a loss of statistical power. However, it is also possible that the
364 effects of solar radiation are stronger in South America than elsewhere. *Liolaemus*, which comprise
365 much of our South American data, span exceptionally large elevational and latitudinal gradients
366 gradient (Pincheira-Donoso, Tregenza, Witt, Hodgson, 2013) and thus may experience exceptional
367 variation in solar radiation; this additional variation could make detecting effects of this variable
368 easier in South America than elsewhere.

369

370 In semi-open habitats, we found no relationship between T_b and wind speed or solar radiation. In
371 these habitats, sun and shade should both be abundant across the landscape. By shuttling in and out of
372 the sun, thermoregulating individuals can maintain body temperatures at, or close to, preferred levels

373 (Huey, 1974), decoupling field-active T_b from incident solar radiation. However, despite this potential
374 for thermoregulation, we still found a positive T_b -air temperature relationship, suggesting that even an
375 abundance of thermoregulation opportunities in these habitats cannot completely degrade the
376 influence of air temperature on body temperature at the global scale. Additionally, lizards inhabiting
377 semi-open habitats may have evolved higher preferred or optimal temperatures than closed forest
378 species (Logan et al., 2013), which could also have contributed to the observed relationship between
379 air temperature and T_b in these environments.

380

381 In contrast to open habitats, relationships between microclimate variables and T_b in closed habitats
382 (forests) did not match predictions of existing biophysical models. Biophysical principles predict that
383 T_b should track air temperature in full shade, as there is no additional heat input from direct solar
384 radiation (Gates, 1980; Campbell & Norman, 1998). As such, many deep forest species are
385 thermoconformers with cooler preferred temperatures (Huey et al., 2009). Yet, we found no
386 relationship between T_b and air temperature in closed habitats. This discrepancy reflects a limitation
387 of current biophysical models, and global microclimate datasets, to accurately capture thermal
388 conditions for ectotherms in these environments. Currently, such datasets, and the biophysical models
389 that use them, must assume a particular shade value (e.g. 100% shade for a closed forest canopy).
390 However, a portion of incident solar radiation still penetrates a closed canopy (Campbell & Norman,
391 1998; also see Eqs 3–5), violating the full shade assumption and allowing T_b to deviate from air
392 temperature. Extending biophysical models to include effects of LAI on solar radiation, as we have
393 done, incorporates this additional source of radiation and eliminates the need to assume a particular
394 shade level, allowing for improved predictions of operative temperatures within forests.

395

396 Incorporating a remotely-sensed measure of canopy LAI greatly improved the R^2 of our regression
397 models in closed forests, indicating that it does capture valuable additional information on thermal
398 environment which is missing from global microclimate variables. Lizards had cooler body
399 temperatures in forests with denser canopies (higher LAI), as predicted by our extended biophysical
400 model. However, we did not find a curvilinear relationship, as predicted, suggesting that there are

401 additional factors that could be usefully incorporated into our model. For example, arboreality will
402 expose species to different thermal conditions from ground level to the canopy (Bakken, 1989;
403 Scheffers et al., 2013; Spicer et al., 2017). Although we attempted to account for the major
404 temperature differences between ground-dwelling and arboreal lizards, more precise data on perch
405 height and type could improve predictions of T_b in treed environments. Also, our model did not
406 include long-wave radiation from the ground that may have been reflected back downwards toward
407 the animal by the canopy (Mahat & Tarboton, 2012), surface conduction; or variation in latent heat
408 loss. Another possibility is that our results reflect both proximate effects of LAI on forest-lizard T_b
409 and adaptive outcomes of living in these environments as sub-canopy lizards in cooler, closed,
410 environments have adapted to prefer, and function at, lower body temperatures (Ruibal, 1961; Hertz,
411 1974; Huey et al., 2009; Munoz et al., 2016). The increase in T_b as LAI decreased is consistent with
412 more solar radiation reaching sub-canopy microenvironments, raising sub-canopy temperatures
413 (Hardwick et al., 2015) and allowing for opportunistic thermoregulation to higher T_b (Otero, Huey,
414 Gorman, 2015), which would permit the occurrence of species adapted to warmer thermal conditions
415 in gaps and edges (Ruibal, 1961; Munoz et al., 2016).

416

417 The reliance of T_b on canopy density in forests highlights the thermal threat of land use and canopy
418 change for ectotherms. Previous biophysical modelling efforts at broad scales have predicted that in a
419 warming world, shade will become increasingly important for ectotherms (Kearney et al., 2009;
420 Sunday et al., 2014). Our results provide empirical evidence of the effect of canopy-shade on body
421 temperature at the scale of these modelled predictions, suggesting that canopy loss would reduce the
422 buffering capacity of shaded environments, further raising body temperatures. In addition to
423 wholesale canopy loss, more subtle changes in canopy density could also raise T_b and narrow thermal
424 safety margins for forest ectotherms, as well as open new opportunities for more warm-adapted
425 species (Huey et al., 2009; but c.f. Logan et al., 2013). In tropical forests, drought causes increased
426 tree mortality along with losses of water content, photosynthetic activity (greenness), and canopy
427 volume (Phillips et al., 2010; Xu et al., 2011; Saatchi et al., 2013; Zhou et al., 2014), leading to
428 increases in the light incident to the forest floor (Slik, 2004). Thus, the increases in drought intensity

429 and/or frequency predicted by global climate models (Dai, 2013; Trenberth et al., 2014) will have
430 indirect effects on T_b via alterations in canopy density, reducing thermal safety margins and the
431 potential for vegetation shade to act as thermal microrefugia (Lenoir et al., 2017) as global
432 temperatures continue to rise. These potential indirect effects of drought highlight the importance of
433 precipitation dynamics, and not just air temperature, for ectotherm thermal niches (Clusella-Trullas et
434 al., 2011).

435

436 Even the relatively coarse-grained, satellite-derived, LAI product used here tripled the fit of T_b -
437 microclimate models in forests, capturing the importance of shade for ectotherms in these
438 environments. At narrower spatial extents, the potential for remote sensing to improve understanding
439 of shade variation and predict thermal habitat quality is even greater. Remote sensing using LiDAR
440 and hyperspectral sensors on airborne platforms capture variation in canopy and sub-canopy
441 vegetation structure at centimeter scales and thus can provide information at scales relevant to
442 individual organisms' movement (George et al., 2015; Frey et al., 2016; Lenoir et al., 2017), and
443 potentially allow for precise estimates of an individual organism's exposure to sun. These products
444 can not only capture the total shade available, but also its spatial configuration, which can have
445 important implications for thermoregulation (Sears & Angilletta, 2015; Sears et al., 2016). Remote
446 sensing data can also improve temporal resolution of predictions of thermal vulnerability in response
447 to land cover change and drought dynamics. Even at the coarser spatial resolutions used here, we have
448 found that canopy structure leaves a predictable signature on lizard body temperatures across the
449 globe, demonstrating the potential of remote sensing products, when properly calibrated, to narrow
450 the predictive envelope of purely biophysical models and provide more precise predictions of T_b in
451 free-living ectotherms across broad spatial extents. Moreover, the sensitivity of T_b to canopy density
452 suggests that changes in forest cover, whether from wholesale land cover change or more subtle
453 alterations of canopy structure, may intensify the thermal challenges faced by organisms in the
454 Anthropocene.

455

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609 **DATA ACCESSIBILITY**

610 All data used in these analyses will be deposited in FigShare or Dryad upon manuscript acceptance.

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Table 1. Consistency of final regression results after randomly sub-setting South American data. N is the number of South American points retained and is equal to the maximum number of points from any other continent. Subsampling and regressions were repeated 1000 times. Sub-setting was not done for closed habitats as only 3 points were outside South America. Columns show the number of regressions coefficients—out of 1000—that matched the sign of the regression on all data, the number of regressions with a significant ($P < 0.05$) coefficient and the means and standard deviations (s.d.) of R^2_m and R^2_c . R^2_m (marginal R^2) is the variance explained by predictors and R^2_c (conditional R^2) is the variance explained by predictors, space and phylogeny. T_b is body temperature, T_{air} is air temperature, SOL is solar radiation, and LAI is the square root of leaf area index.

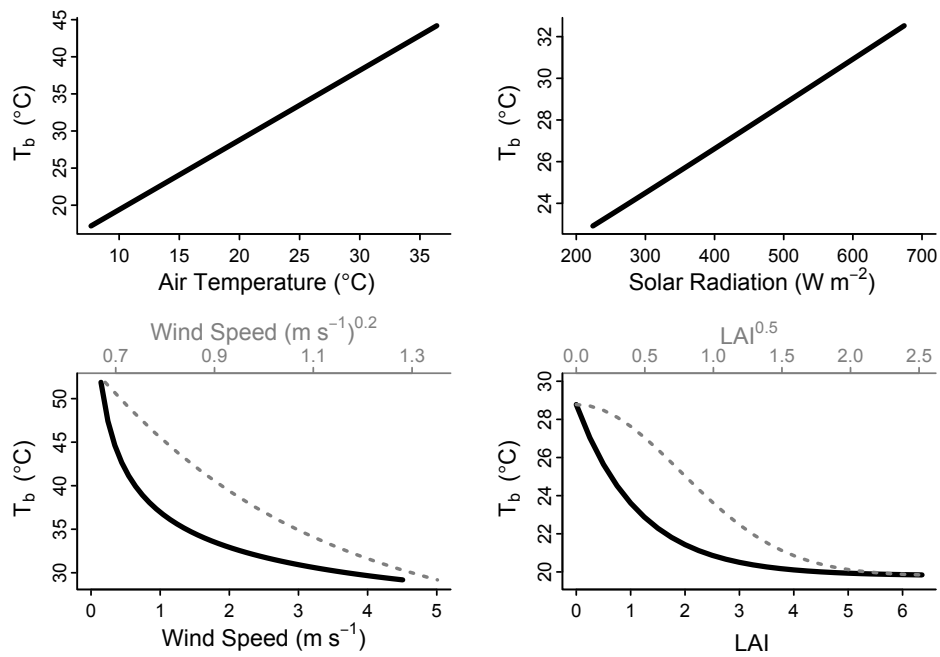
Habitat	Model	Tair		SOL		R^2_m		R^2_c		N
		+	P<0.05	+	P<0.05	mean	s.d.	mean	s.d.	
All	$T_b \sim T_{air} + SOL$	1000	1000	447	2	0.16	0.04	0.45	0.26	28
Semi	$T_b \sim T_{air}$	1000	810			0.08	0.04	0.39	0.33	20
Open	$T_b \sim T_{air} + SOL + WS$	1000	1000	970	113	0.40	0.06	0.57	0.18	11

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621 **Fig. 1.** Predicted relationships between body temperature (T_b), microclimate, and leaf area index
622 (LAI). Predictions are based on a biophysical model of body temperature for a non-thermoregulating
623 lizard (Eq. 9). Parameters for model predictions are given in Appendix 1 and Table S1.1.
624 Relationships with air temperature, solar radiation and wind speed were modelled in full sun (LAI=0).
625 The grey axes and lines for wind speed and LAI show predicted relationships after a fifth root
626 transformation of wind speed and a square root transformation of LAI to allow comparison with
627 patterns in Fig. 4.

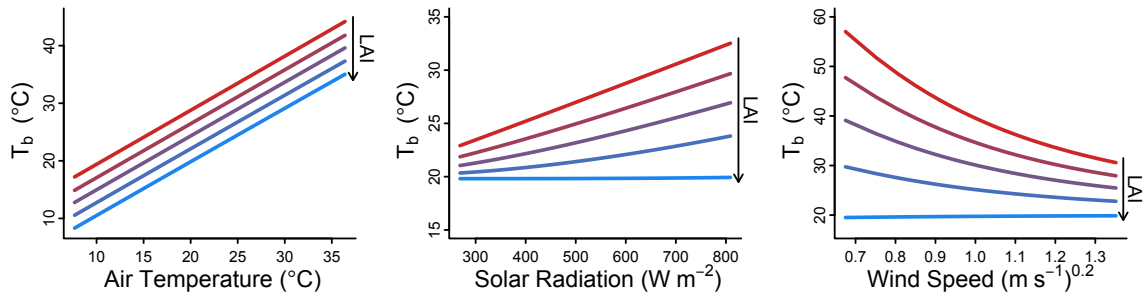
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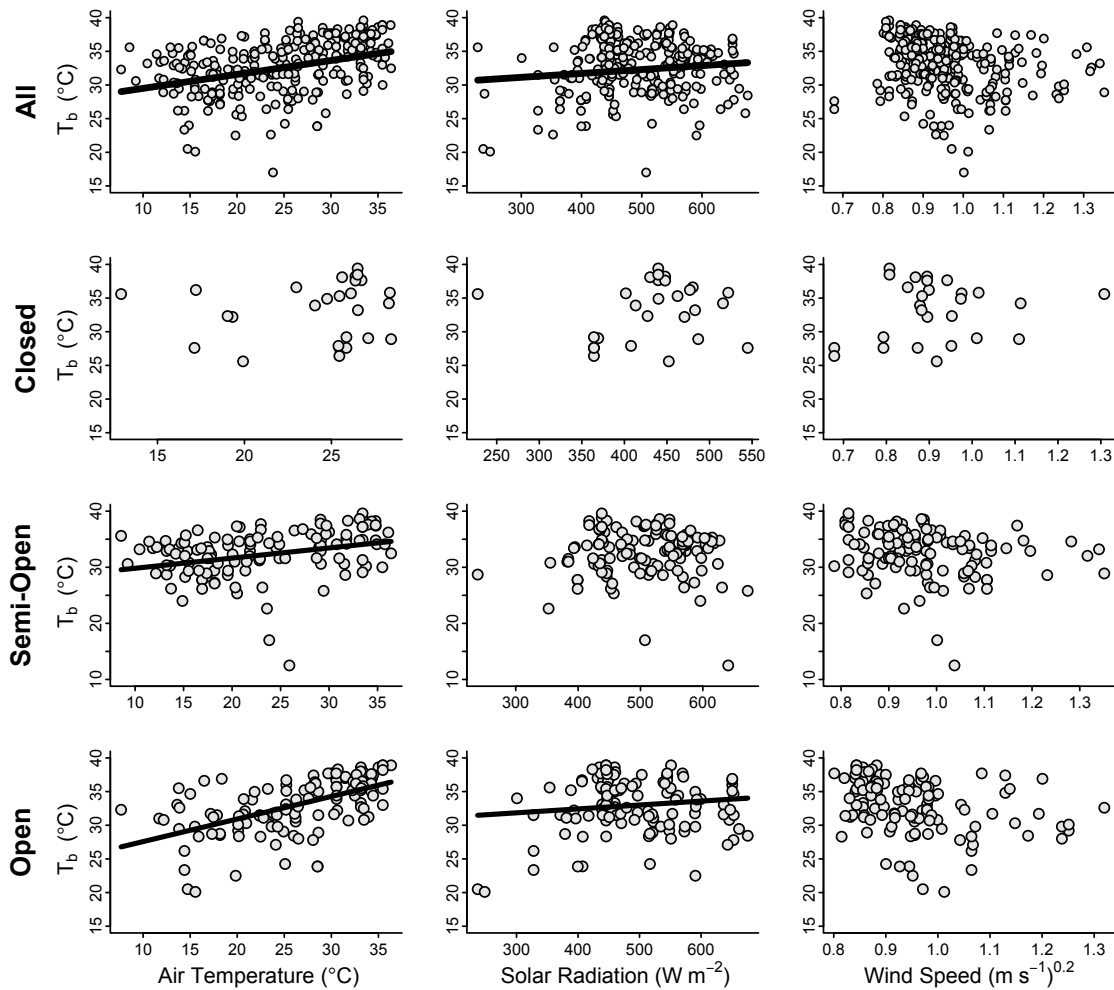
634

635 **Fig. 2.** Predicted effect of LAI on the body temperature (T_b) – microclimate relationship for a non-
636 thermoregulating lizard at equilibrium with its environment. The arrows depict direction of increasing
637 LAI, from low (red) to high (blue). Relationships were modelled using Eq. 9 and all parameters are
638 given in Appendix S1 and Table S1.1. LAI is predicted to have an interactive effect on T_b with solar
639 radiation and wind speed but not air temperature.

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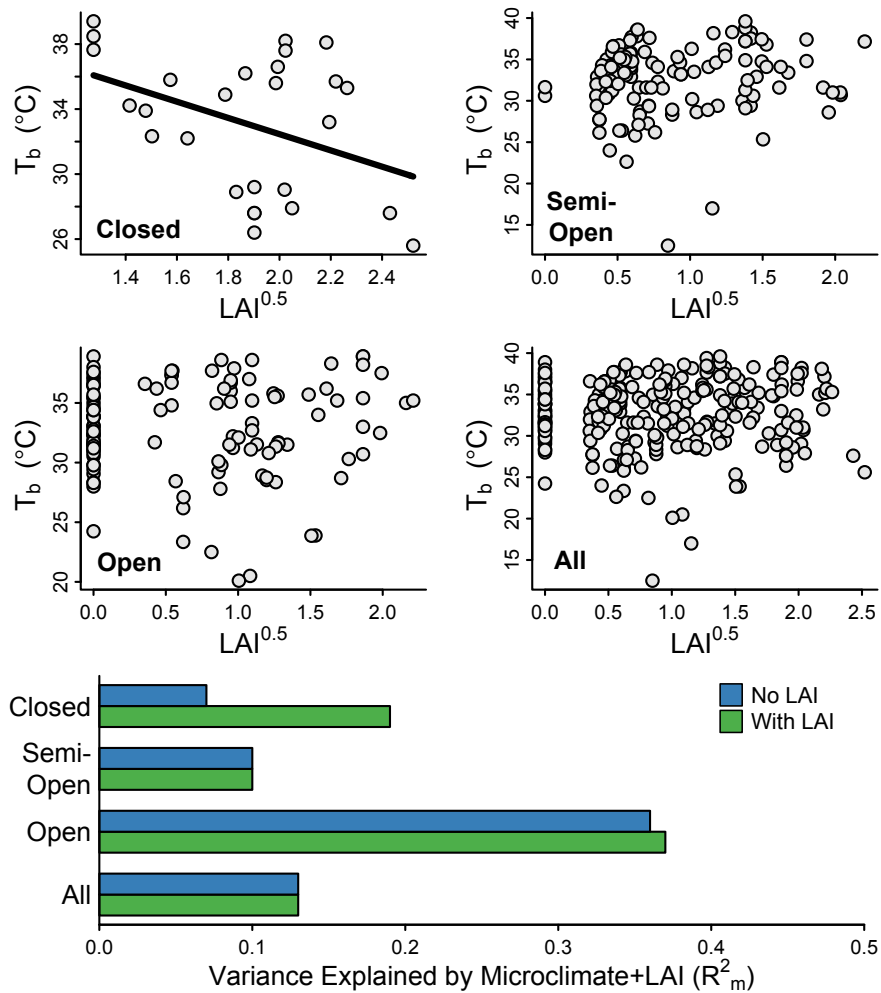
645 **Fig. 3.** Relationship between body temperature (T_b) and microclimate for lizards in all habitats and
646 divided by major habitat types that vary in shade availability. Microclimate conditions are daytime
647 averages across the months of T_b collection. Significant ($P < 0.05$) relationships after accounting for
648 spatial and phylogenetic relationships are shown as black regression lines. The variance in T_b
649 explained by microclimate (with linear wind speed) was lowest in closed (0.07%, $n=27$) and semi-
650 open habitats (11%, $n=123$), highest in open habitats (36%, $n=119$), and low in the combined data
651 (0.13, $n=269$). Wind speed was fifth root transformed to reduce skew and leverage of extreme points.

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Fig. 4. Relationship between lizard body temperature (T_b) and leaf area index (LAI) in closed, semi-open, and open habitats and all habitats pooled. The relationship was significant ($P=0.03$) in closed (forest) habitats and was in the direction (negative) predicted by theory, though not curvilinear (see Fig. 1). The bottom panel shows the unique variance explained by environment predictors (marginal R^2) without and with LAI. LAI had little effect on model fit apart from in closed forests.

670 **APPENDIX 1 – Data Sources**

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937

SUPPLEMENTARY INFORMATION

REMOTE SENSING RESTORES PREDICTABILITY OF ECTOTHERM BODY TEMPERATURE IN THE WORLD'S FORESTS

APPENDIX S1: SUPPLEMENTARY METHODS, TABLES AND FIGURES

Supplementary Methods

Modelling operative temperature (T_e) in full sun

We derived the expected relationships between T_b of a non-thermoregulating lizard at equilibrium with its environment and air temperature, solar radiation, wind speed and shade level by modelling T_e for a theoretical lizard on a flat surface and varying each of these variables individually or jointly, assuming $T_b = T_e$ for a non-thermoregulating lizard at equilibrium. Initial parameters for these calculations are given in Table S1.1. Solar radiation was manipulated by varying latitude while controlling for time of day and year.

We modelled T_e on a flat surface following Buckley (2007) and Sears et al. (2011), using the equation:

$$T_e = T_{air} + \frac{R_{solar} + R_{lw} - \varepsilon_s \sigma (T_{air} + 273)^4}{4\sigma (T_{air} + 273)^3 + c_p \left(1.4 + 0.135 \sqrt{\frac{v}{d}}\right)} \quad \text{Eq S1}$$

where T_{air} is air temperature in degrees Celsius, R_{solar} is absorbed incoming solar (short-wave) radiation, R_{lw} is absorbed long-wave radiation, ε_s is animal emissivity which we set at 0.965, following Buckley (2007), σ is the Stefan-Boltzmann constant ($5.67 \times 10^{-8} \text{ W m}^{-2} \text{ K}^{-4}$; Buckley 2007), c_p is the specific heat of air ($29.3 \text{ J mol}^{-1} \text{ K}^{-1}$) (Buckley, 2007), d is the characteristic dimension of the animal, and v is wind velocity. We set d equal to the snout-vent length of our theoretical lizard, which assumes the animal is parallel to wind direction (Campbell & Norman, 1998).

We calculated the absorbed long-wave radiation following Buckley's (2007) equation A23:

$$R_{lw} = \alpha_L (F_a L_a + F_g L_g) \quad \text{Eq S2}$$

α_L is absorptivity in the long-wave (thermal) waveband, set to 0.965, following Buckley (2007). F_a and F_g are view factors for long-wave radiation from the air and ground, respectively, both set to 0.5 (Buckley, 2007; Sears *et al.*, 2011). L_a is the long-wave radiation from the air, calculated using Buckley's (2007) Eq A21:

973

$$L_a = \varepsilon_{ac} \sigma (T_{air} + 273)^4 \quad \text{Eq S3}$$

975

976 ε_{ac} is clear-sky emissivity which was calculated using Buckley's (2007) equation A12:

977

$$\varepsilon_{ac} = 9.2 \times 10^{-6} (T_{air} + 273)^2 \quad \text{Eq S4}$$

979

980 L_g in Eq S2 is the long-wave radiation from the ground and was calculated following Buckley's
981 (2007) equation A22:

982

$$L_g = \varepsilon_s \sigma (T_s + 273)^4 \quad \text{Eq S3}$$

984

985 where T_s is the ground surface temperature in degrees Celsius and ε_s is the emissivity of the ground.
986 According to Campbell and Norman (1998), emissivity is between 0.95 and 1.0 for most natural
987 surfaces, so we use 0.965, which matches the emissivity of our theoretical lizard.

988

989 Solar (short-wave) radiation absorbed by an animal in full sun was modelled following Buckley's
990 (2007) equation A23 (with slightly different notation):

991

$$R_{solar} = \alpha_s (F_p S_p + F_d S_d + F_r S_r) \quad \text{Eq S4}$$

993

994 In Eq S4, α_s is the absorptivity of solar radiation, set to 0.9 for lizards (Gates, 1980; Buckley, 2007).
995 F_p , F_d , and F_r are view factors between the lizard and direct solar radiation (S_p), diffuse solar radiation
996 (S_d), and reflected solar radiation (S_r), respectively. F_r was set to 0.5 (Buckley, 2007; Sears *et al.*,
997 2011), as was F_d (Sears *et al.*, 2011). Buckley (2007) set F_d to 0.8, but this slight difference would not
998 affect our conclusions about the shape and direction of the relationships between T_e and microclimate.
999 F_p was modelled following Sears *et al.*'s (2011) Equation 5 and assumes the lizard is a cylinder with
1000 rounded ends:

$$F_p = 1 + \frac{4h \sin \Theta}{4 + \frac{\pi w}{w}} \quad \text{Eq S5}$$

1002

1003 where h is the SVL of the lizard (length of cylinder) and w is the body width (diameter of cylinder). Θ
1004 is the angle between the solar beam and the animal's longitudinal axis, which we assumed to be 90° .

1005

1006 Direct solar radiation reaching the Earth's surface (S_p) in full sun was calculated following Sears *et al.*
1007 *et al.*'s (2011) equations 6 and 9:

1008

1009
$$S_p = \bar{S}_0 \left(1 + 2 \times \cos \left(\frac{2\pi}{365} J \right) \right) \tau^m \cos z$$
 Eq S6

1010

1011 where \bar{S}_0 is the solar constant (1360 W m^{-2}), J Julian day, z is the zenith angle, τ is the optical
1012 transmittance of the atmosphere. τ values between 0.6 and 0.7 are typical of clear days (Gates, 1980;
1013 Campbell & Norman, 1998), so we set $\tau=0.65$. m is the optical air mass number, given by Sears et al.
1014 (2011), following Campbell & Norman (1998), as:

1015

1016
$$m = \frac{101.3 e^{\frac{a}{8200}}}{101.3 \cos z} = \frac{e^{\frac{a}{8200}}}{\cos z}$$
 Eq S7

1017

1018 where a is elevation in meters above sea level. We calculated zenith angle, assuming a flat surface
1019 with no surrounding topography, as a function of latitude, longitude and hour following Sears et al.
1020 (2011) and Campbell & Norman (1998):

1021

1022
$$z = \cos^{-1}(\sin \Phi \sin \delta + \cos \Phi \cos \delta \cos h)$$
 Eq S8

1023

1024 where Φ is latitude, δ is solar declination, and h is the hour angle of the sun. We calculated
1025 declination using Campbell & Norman's(1998) equation 11.2:

1026

1027
$$\delta = 0.39785 \sin(278.97 + 0.9856J + 1.9165 \sin(356.6 + 0.9856J))$$
 Eq S9

1028

1029 We calculate hour angle of the sun as:

1030
$$h = 15(t - 12 + L_{cor} + E_T)$$
 Eq S10

1031

1032 where t is time of day, L_{cor} is a longitudinal correction and E_T is the time equation, calculated from
1033 Sears et al. (2011) and Campbell & Norman (1998) as:

1034

1035
$$E_T = 3600^{-1}(-104.7 \sin f + 596.2 \sin 2f + 4.3 \sin 3f - 12.7 \sin 4f$$

1036
$$-429.3 \cos f - 2.0 \cos 2f + 19.3 \cos 3f)$$
 Eq S11

1037

1038 where $f = 279.575 + 0.9856J$. The longitudinal correction (L_{cor}) is the longitude plus 4 minutes for
1039 each degree east of a standard meridian (minus 4 minutes for each degree west), where standard
1040 meridians are located at $0^\circ, 15^\circ, 30^\circ, \dots, 345^\circ$ (Campbell & Norman, 1998).

1041

1042 Diffuse solar radiation (S_d) was modelled following Buckley (2007) and Campbell and Norman
 1043 (1998):

1044

$$1045 \quad S_d = 0.3\bar{S}_0(1 - \tau^m) \cos z \quad \text{Eq S10}$$

1046

1047 Reflected solar radiation (S_r) was modelled as a function of surface (ground) albedo (α) and the direct
 1048 and diffuse solar radiation (Campbell & Norman, 1998; Buckley, 2007; Sears *et al.*, 2011):

1049

$$1050 \quad S_r = \alpha(S_p + S_d) \quad \text{Eq S10}$$

1051

1052 Modelling operative temperature (T_e) under the forest canopy

1053

1054 Shade has previously been incorporated into T_e models by reducing the amount of solar radiation
 1055 reaching an animal:

1056

$$1057 \quad T_e = T_{air} + \frac{(1 - S)R_{solar} + R_{lw} - \epsilon_s \sigma (T_{air} + 273)^4}{4\sigma (T_{air} + 273)^3 + c_p \left(1.4 + 0.135 \sqrt{\frac{v}{d}}\right)} \quad \text{Eq S12}$$

1058 where S is the proportion of an animal in the shade (Sears *et al.*, 2011). However, shade imposed by a
 1059 forest canopy will affect both solar radiation and long-wave radiation (Campbell, 1986; Campbell &
 1060 Norman, 1998; Webster *et al.*, 2017). The direct solar radiation reaching an animal in full sun (S_p) was
 1061 modelled using Equation S4. However, under a forest canopy, only a proportion of the potential direct
 1062 solar radiation will penetrate the canopy. We modelled the direct solar radiation reaching below the
 1063 canopy ($S_{p,sub}$) following Campbell and Norman (1998):

1064

$$1065 \quad S_{p,sub} = \omega_p S_p \quad \text{Eq S13}$$

1066

1067 where ω_p is the proportion of direct solar radiation that makes it through the canopy, which is an
 1068 exponential function of LAI (Campbell & Norman, 1998):

1069

$$1070 \quad \omega_p = \exp(-\sqrt{\alpha_c} K_{b,z} LAI) \quad \text{Eq S14}$$

1071

1072 where α_c is the average absorptivity of the canopy, set to 0.8 (Page 248 in Campbell & Norman,
 1073 1998), $K_{b,z}$ is the extinction coefficient for direct solar radiation at zenith angle z . $K_{b,z}$ was modelled as
 1074 a function of z , assuming a spherical leaf angle distribution (Campbell & Norman, 1998):

1075

1076
$$K_{b,z} = \frac{\sqrt{1 + \tan^2 z}}{2.00132}$$
 Eq S15

1077

1078 Diffuse solar radiation below the canopy ($S_{d,sub}$) was found by numerically integrating Equation 15.5
1079 in Campbell and Norman (1998):

1080

1081
$$S_{d,sub} = S_d 2 \int_0^{\frac{\pi}{2}} \exp(-\sqrt{\alpha_c} K_{b,z} LAI) \sin z \cos z dz$$
 Eq S16

1082

1083 We then modelled the solar radiation reflected from the ground under the canopy ($S_{r,sub}$) as:

1084

1085
$$S_{r,sub} = \alpha (S_{p,sub} + S_{d,sub})$$
 Eq S17

1086

1087 and total solar radiation reaching the animal under the canopy ($R_{solar,sub}$) as:

1088

1089
$$R_{solar,sub} = \alpha_s (F_p S_{p,sub} + F_d S_{d,sub} + F_r S_{r,sub})$$
 Eq S18

1090

1091 The forest canopy will also affect the long-wave radiation incident on an animal if the canopy is a
1092 different temperature than the air (Webster *et al.*, 2017). We modelled the below-canopy long-wave
1093 radiation ($L_{a,sub}$) following Webster *et al.* (2017), but accounting for canopy emissivity (ϵ_c):

1094

1095
$$L_{a,sub} = V_s L_a + (1 - V_s) \epsilon_c \sigma (T_{can} + 273)^4$$
 Eq S19

1096

1097 where T_{can} is canopy temperature in °C, V_s is a view factor delineating proportion of radiation comes
1098 from clear sky and ϵ_c is set to 0.99 (Page 273 in Campbell & Norman, 1998). We modelled T_{can} using
1099 the empirical relationship derived by Webster *et al.* (2017):

1100

1101
$$T_{can} = 2.36 + 0.88 T_{air} + 0.0073 (S_{p,sub} + S_{d,sub} + S_{r,sub})$$
 Eq S20

1102

1103 V_s varies between 0 and 1, where 0 indicates complete canopy cover and 1 no canopy cover. We
1104 modelled V_s as an exponential function of LAI. When LAI is zero, V_s is 1.0 as all long-wave radiation
1105 from the air comes from the sky (there is no canopy). As LAI increases leaves block the sky,
1106 decreasing V_s . The proportion of direct (beam) solar radiation that penetrates the canopy (ω_p) is given
1107 in Eq S14. This reflects value reflects the amount of direct light reaching the ground between and
1108 through leaves (as determined by α_c , canopy absorptivity). Thus, we modelled V_s based on Eq S14,
1109 assuming black leaves ($\alpha_c=1.0$):

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$$V_s = \exp(-K_{b,z}LAI) \quad \text{Eq S14}$$

We used these equations to model the expected relationships between air temperature, solar radiation, LAI and wind speed. Parameters used in models, but not defined above are given in Table S1.1.

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Table S1.1. Parameter values not given in the supplementary text above for modelling operative temperature of a lizard on a flat surface. Air temperature, wind speed, shade level and latitude (as a proxy for solar radiation) were varied between the given ranges to derive predicted relationships with lizard body temperature. Variable ranges were based on ranges in our global data set

Variable	Default	Range
Air Temperature (°C)	20	7.7–36.4
Substrate Temperature (°C)	20	–
Wind speed (m s ⁻¹)	5	0.14–4.51
Latitude (DD)	34.0	25.5–47.5
Longitude (DD)	-111.0	–
Hour (local time)	11:00	–
Julian Day	10	–
LAI	0	0–6.4
Surface albedo	0.2	–
Elevation (m.a.s.l.)	100	–
Snout-Vent Length (cm)	10	–
Body Diameter (cm)	2	–

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Table S1.2. Reclassification of land cover categories in Tuanmu & Jetz (2014) into closed, semi-open and open habitat types.

Class	Description	Habitat type
1	Evergreen/Deciduous Needleleaf Trees	Closed
2	Evergreen Broadleaf Trees	Closed
3	Deciduous Broadleaf Trees	Closed
4	Mixed/Other Trees	Semi-Open
5	Shrubs	Semi-Open
6	Herbaceous Vegetation	Open
7	Cultivated and Managed Vegetation	Open
8	Regularly Flooded Vegetation	Open
9	Urban/Built-up	Open
10	Snow/Ice	-
11	Barren	Open
12	Open Water	-

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Table S1.3. Regression of T_b on microclimate and canopy structure for global data and all habitats. For each variable the sign of the coefficient and P-value are given. $P < 0.05$ are in bold. All regressions accounted for spatial and phylogenetic autocorrelation. R_m^2 (marginal R^2) is the variance explained by predictors and R_c^2 (conditional R^2) is the variance explained by predictors, space and phylogeny. T_b is body temperature, T_{air} is air temperature, SOL is solar radiation, WS is the fifth root of wind speed and LAI is the square root of leaf area index.

Model	T_{air}	SOL	WS	WS^2	LAI	LAI^2	LAI×SOL	LAI×WS	R_m^2	R_c^2
No LAI										
$T_b \sim T_{air} + SOL + WS + WS^2$	+, 3e-6	+, 0.01	-, 0.54	+, 0.61					0.13	0.42
$T_b \sim T_{air} + SOL + WS$	+, 2e-6	+, 0.01	-, 0.27						0.13	0.42
$T_b \sim T_{air} + SOL$	+, 7e-10	+, 0.02							0.13	0.38
With LAI										
$T_b \sim T_{air} + SOL + WS + WS^2 + LAI + LAI^2 + LAI \times SOL + LAI \times WS$	+, 6e-6	+, 0.34	-, 0.36	+, 0.44	-, 0.27	+, 0.42	+, 0.49	+, 0.50	0.14	0.43
$T_b \sim T_{air} + SOL + WS + WS^2 + LAI + LAI^2 + LAI \times SOL$	+, 5e-6	+, 0.35	-, 0.46	+, 0.52	-, 0.37	+, 0.44	+, 0.50		0.14	0.42
$T_b \sim T_{air} + SOL + WS + WS^2 + LAI + LAI^2$	+, 2e-6	+, 0.02	-, 0.51	+, 0.57	-, 0.46	+, 0.54			0.14	0.40
$T_b \sim T_{air} + SOL + WS + WS^2 + LAI$	+, 3e-6	+, 0.02	-, 0.50	+, 0.57	-, 0.64				0.13	0.42
$T_b \sim T_{air} + SOL + WS + LAI$	+, 2e-6	+, 0.03	-, 0.27		-, 0.69				0.13	0.41

Table S1.4. Results of final regressions of T_b on microclimate and canopy structure using a Pagel's lambda transformed phylogenetic variance-covariance matrix. For each variable the sign of the coefficient and P-value are given. $P < 0.05$ are in bold. All regressions accounted for spatial and phylogenetic autocorrelation. R^2_m (marginal R^2) is the variance explained by predictors and R^2_c (conditional R^2) is the variance explained by predictors, space and phylogeny. T_b is body temperature, T_{air} is air temperature, SOL is solar radiation, WS is the fifth root of wind speed and LAI is the square root leaf area index.

Model		T_{air}	SOL	WS	LAI	R^2_m	R^2_c
All	$T_b \sim T_{air} + SOL$	+ ; 6e-10	+ ; 0.02			0.13	0.40
Closed	$T_b \sim LAI$				- ; 0.03	0.15	0.17
Semi	$T_b \sim T_{air}$	+ ; 3e-4				0.10	0.15
Open	$T_b \sim T_{air} + SOL$	+ ; 1e-15	+ ; 4e-3			0.35	0.52

Table S1.5. Regression of T_b on microclimate and canopy structure for open habitats in the global data. For each variable the sign of the coefficient and P-value are given. $P < 0.05$ are in bold. All regressions accounted for spatial and phylogenetic autocorrelation. R^2_m (marginal R^2) is the variance explained by predictors and R^2_c (conditional R^2) is the variance explained by predictors, space and phylogeny. T_b is body temperature, T_{air} is air temperature, SOL is solar radiation, WS is the fifth root of wind speed and LAI is the square root leaf area index.

Model	T_{air}	SOL	WS	WS^2	LAI	LAI^2	LAI×SOL	LAI×WS	R^2_m	R^2_c
No LAI										
$T_b \sim T_{air} + SOL + WS + WS^2$	+, 2e-7	+, 5e-4	-, 0.08	+, 0.10					0.37	0.64
$T_b \sim T_{air} + SOL + WS$	+, 2e-9	+, 0.001	-, 0.11						0.36	0.58
$T_b \sim T_{air} + SOL$	+, 1e-15	+, 0.004							0.35	0.53
With LAI										
$T_b \sim T_{air} + SOL + WS + WS^2 + LAI + LAI^2 + LAI \times SOL + LAI \times WS$	+, 1e-6	+, 0.38	-, 0.16	+, 0.17	-, 0.64	+, 0.09	+, 0.12	-, 0.27	0.40	0.64
$T_b \sim T_{air} + SOL + WS + WS^2 + LAI + LAI^2 + LAI \times SOL$	+, 7e-7	+, 0.26	-, 0.07	+, 0.05	-, 0.07	+, 0.08	+, 0.19		0.40	0.66
$T_b \sim T_{air} + SOL + WS + WS^2 + LAI + LAI^2$	+, 5e-8	+, 0.002	-, 0.08	+, 0.10	-, 0.08	+, 0.17			0.39	0.64
$T_b \sim T_{air} + SOL + WS + WS^2 + LAI$	+, 4e-9	+, 0.002	-, 0.08	+, 0.11	-, 0.21				0.38	0.65

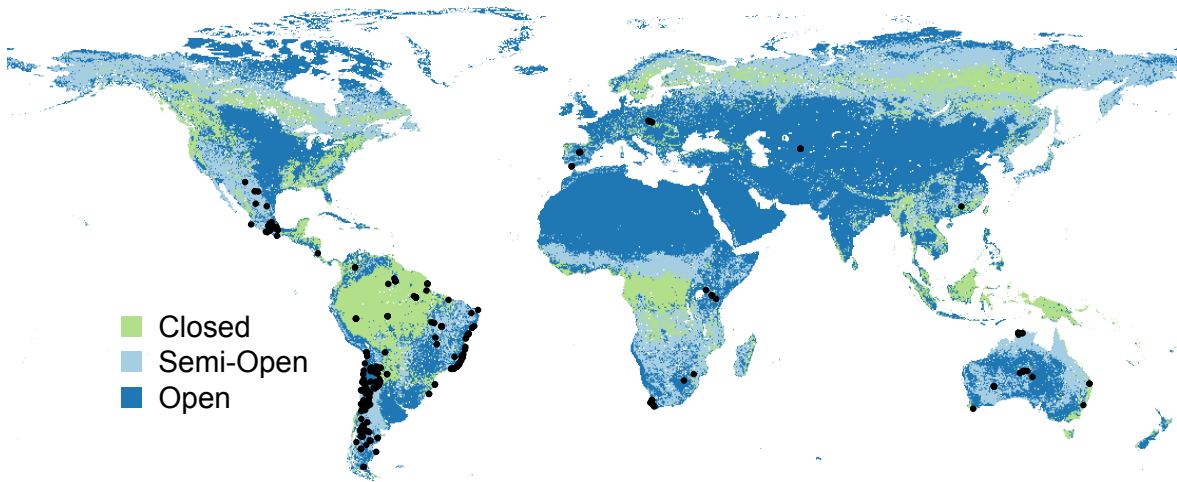
Table S1.6. Regression of T_b on microclimate and canopy structure for semi-open habitats in the global data. For each variable the sign of the coefficient and P-value are given. $P < 0.05$ are in bold. All regressions accounted for spatial and phylogenetic autocorrelation. R_m^2 (marginal R^2) is the variance explained by predictors and R_c^2 (conditional R^2) is the variance explained by predictors, space and phylogeny. T_b is body temperature, T_{air} is air temperature, SOL is solar radiation, WS is the fifth root of wind speed and LAI is the square root leaf area index.

Model	T_{air}	SOL	WS	WS^2	LAI	LAI^2	LAI×SOL	LAI×WS	R_m^2	R_c^2
No LAI										
$T_b \sim T_{air} + SOL + WS + WS^2$	+ ; 0.01	+; 0.70	−; 0.36	+; 0.38					0.11	0.13
$T_b \sim T_{air} + SOL + WS$	+ ; 0.004	+; 0.85	−; 0.71						0.10	0.14
$T_b \sim T_{air} + WS$	+ ; 0.004		−; 0.73						0.10	0.14
$T_b \sim T_{air}$	+ ; 3e-4								0.10	0.15
With LAI										
$T_b \sim T_{air} + SOL + WS + WS^2 + LAI + LAI^2 + LAI \times SOL + LAI \times WS$	+ ; 0.01	+; 0.72	−; 0.39	+; 0.37	+; 0.89	+; 0.59	−; 0.75	−; 0.80	0.11	0.19
$T_b \sim T_{air} + SOL + WS + WS^2 + LAI + LAI^2 + LAI \times SOL$	+ ; 0.009	+; 0.71	−; 0.35	+; 0.35	−; 1.0	+; 0.60	−; 0.72		0.11	0.19
$T_b \sim T_{air} + SOL + WS + WS^2 + LAI + LAI^2$	+ ; 0.009	+; 0.89	−; 0.31	+; 0.32	−; 0.40	+; 0.45			0.11	0.18
$T_b \sim T_{air} + SOL + WS + WS^2 + LAI$	+ ; 0.01	+; 0.83	−; 0.33	+; 0.35	−; 0.69				0.11	0.14
$T_b \sim T_{air} + SOL + WS + LAI$	+ ; 0.006	+; 0.92	−; 0.71		−; 0.86				0.10	0.15
$T_b \sim T_{air} + WS + LAI$	+ ; 0.004		−; 0.72		−; 0.86				0.10	0.15

Table S1.7. Regression of T_b on microclimate and canopy structure for closed habitats in the global data. For each variable the sign of the coefficient and P-value are given. $P < 0.05$ are in bold. All regressions accounted for spatial and phylogenetic autocorrelation. R^2_m (marginal R^2) is the variance explained by predictors and R^2_c (conditional R^2) is the variance explained by predictors, space and phylogeny. T_b is body temperature, T_{air} is air temperature, SOL is solar radiation, WS is the fifth root of wind speed and LAI is the square root leaf area index.

Model	T_{air}	SOL	WS	WS^2	LAI	LAI^2	LAI×SOL	LAI×WS	R^2_m	R^2_c
No LAI										
$T_b \sim T_{air} + SOL + WS + WS^2$	+; 0.60	+; 0.80	+; 0.58	−; 0.65					0.08	0.09
$T_b \sim T_{air} + SOL + WS$	+; 0.54	+; 0.44	+; 0.29						0.07	0.12
$T_b \sim SOL + WS$		+; 0.35	+; 0.35						0.06	0.13
$T_b \sim WS$			+; 0.38						0.03	0.07
With LAI										
$T_b \sim T_{air} + SOL + WS + WS^2 + LAI + LAI^2 + LAI \times SOL + LAI \times WS$	−; 0.91	+; 0.32	−; 0.47	−; 0.67	−; 0.53	−; 0.88	−; 0.39	+; 0.08	0.28	0.28
$T_b \sim T_{air} + SOL + WS + WS^2 + LAI + LAI^2 + LAI \times WS$	+; 0.75	+; 0.34	−; 0.52	−; 0.72	−0.16	−; 0.71		+; 0.10	0.27	0.27
$T_b \sim T_{air} + SOL + WS + WS^2 + LAI + LAI^2$	+; 1.0	+; 0.95	+; 0.42	−; 0.47	−; 0.73	−; 0.89			0.20	0.20
$T_b \sim T_{air} + SOL + WS + WS^2 + LAI$	−; 0.98	+; 0.92	+; 0.40	−; 0.45	−; 0.04				0.20	0.20
$T_b \sim T_{air} + SOL + WS + LAI$	+; 0.90	+; 0.40	+; 0.32		−; 0.05				0.19	0.19
$T_b \sim SOL + WS + LAI$		+; 0.37	+; 0.31		−; 0.03				0.20	0.20
$T_b \sim WS + LAI$			+; 0.35		−; 0.03				0.18	0.18
$T_b \sim LAI$					−; 0.03				0.15	0.17

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Figure S1.1. Location of lizard populations (n=269) with T_b data, overlaid over our habitat classification.