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# Climatic Predictors of Temperature Performance Curve Parameters in Ectotherms Imply Complex Responses to Climate Change

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ABSTRACT: Determining organismal responses to climate change is one of biology's greatest challenges. Recent forecasts for future climates emphasize altered temperature variation and precipitation, but most studies of animals have largely focused on forecasting the outcome of changes in mean temperature. Theory suggests that extreme thermal variation and precipitation will influence species performance and hence affect their response to changes in climate. Using an information-theoretic approach, we show that in squamate ectotherms (mostly lizards and snakes), two fitness-influencing components of performance, the critical thermal maximum and the thermal optimum, are more closely related to temperature variation and to precipitation, respectively, than they are to mean thermal conditions. By contrast, critical thermal minimum is related to mean annual temperature. Our results suggest that temperature variation and precipitation regimes have had a strong influence on the evolution of ectotherm performance, so that forecasts for animal responses to climate change will have to incorporate these factors and not only changes in average temperature.

*Keywords:* reptiles, species distributions, reaction norm, operative temperature, arid zones, thermoregulation.

#### Introduction

How organisms respond to ongoing anthropogenic climate change and what conservation actions should be taken to assist species to respond are among the most significant and controversial topics in biology today (Hoegh-Guldberg et al. 2008; Williams et al. 2009). While it is widely appreciated that increases in temperature have significantly affected species ranges and their phenologies (e.g., Stachowicz et al. 2002; Forcada et al. 2006; Parmesan 2006), several unexpected outcomes of change have also been

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identified, including substantial current and forecast impacts on tropical organisms (Pounds et al. 2006; Deutsch et al. 2008; Raxworthy et al. 2008; Chen et al. 2009; Huey et al. 2009; Sinervo et al. 2010).

Much of this research, at least for ectotherms, has been concerned with the responses of species to changing average thermal conditions (mostly, mean maximum temperature or mean ambient temperature conditions) or to the altered timing of particular events (e.g., Bradshaw and Holzapfel 2006; Deutsch et al. 2008; but see Kearney et al. 2009, McKechnie and Wolf 2010). However, major components of current and forecast climate change include an increase in the frequency of extreme events, alteration of the diurnal temperature range, changes in precipitation and cloud cover, and interactions between temperature and precipitation change (Vose et al. 2005; Solomon et al. 2007; Adler et al. 2008; Zhou et al. 2009b). Changes in precipitation and cloud cover are more complex and spatially variable than those in temperature conditions. Nonetheless, expected rainfall trends include an increase in the frequency of large events and a decline in predictability (Jentsch et al. 2007; Knapp et al. 2008), an increase in tropical and high-latitude rainfall, a decline at midlatitudes, and substantial variation between hemispheres (Meehl et al. 2007; Trenberth et al. 2007; Adler et al. 2008; Zhou et al. 2009b). Changes in cloud-base height and fog in several coastal and montane areas are also predicted (Nair et al. 2003; Williams et al. 2008; Fischer et al. 2009; Rovito et al. 2009). How these changes are likely to influence organismal responses has not been clearly resolved, although interactions between precipitation and temperature change and changes in their variability are likely to be significant (e.g., Bonebrake and Mastrandrea 2010).

Ectotherms are the predominant group of terrestrial animal species that either largely comply with their surrounding thermal environment or rely mostly on behav-

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ioral thermoregulation for optimum performance, fitness maximization, and ultimately, survival (Angilletta 2009). Several lines of evidence suggest that the performance of most animals should therefore be affected by temperature variation, precipitation, and cloud cover. First, extreme events may impose strong selection that can lead to substantial population change (for theory, see Endler 1986; Hoffman and Parsons 1997; for empirical support, see Grant and Grant 1993; Brown and Brown 1998; Parmesan et al. 2000; Gutschick and BassiriRad 2003; Jentsch et al. 2007; Parker et al. 2008; Wegner et al. 2008; Welbergen et al. 2008; Pelini et al. 2009).

Second, from a theoretical perspective, thermal tolerance breadth should increase in more variable environments to optimize fitness (Janzen 1967; Stevens 1989; see also Snyder and Weathers 1975; Addo-Bediako et al. 2000; Ghalambor et al. 2006). However, predictions from optimality models of thermal performance curves (Lynch and Gabriel 1987; Gilchrist 1995) have received mixed empirical support (Angilletta 2009) and have been little explored with regard to precipitation variability. Bonebrake and Mastrandrea (2010) have recently applied the optimality model of Lynch and Gabriel (1987) to predict relative changes in fitness given variation in both temperature and precipitation across the globe. Their results indicate that changes in precipitation variability significantly influence performance curves and relative fitness across latitudes, with several outcomes differing substantially from those of models based on temperature alone.

Third, the range and frequency of operative temperatures  $(T_{a})$  that integrate not just ambient temperature but multiple energy paths (radiation, convection, transpiration, conduction; Gates 1980) within the habitat influence the thermoregulatory opportunities of ectotherms and thus likely affect the fitness of these organisms. The effects of precipitation on performance curves can result from either a direct effect on the availability of optimal  $T_e$  for thermoregulation (Avery 1971; Kingsbury 1994; Marquis et al. 2008; Clusella-Trullas et al. 2009) or a direct effect on hydration status, or both. For example, selected body temperatures can be lower in arid areas with high dehydration risks (Lorenzon et al. 1999; Hertz 1992). Finally, precipitation could also have an indirect effect on temperature performance curves by affecting resource availability and thus altering the potential amount of energy intake of reptiles (Dunham 1978; Tinkle et al. 1993; Lorenzon et al. 2001). In productive areas, faster rates of digestion from higher preferred body temperatures may provide increased food consumption, growth rate, and predator avoidance (e.g., Harlow et al. 1976; Blouin-Demers and Weatherhead 2001), whereas food deprivation in poor resource areas may result in lower selected body temperatures to minimize metabolic costs (Brown and Griffin 2005; Wang et al. 2006).

Given the potential significance of temperature and precipitation changes for ectotherm fitness, covariation between these environmental variables and the major parameters of ectotherm performance curves, such as critical temperature limits and optimal temperature, should be pronounced. If occasional high-temperature events have had a significant role in shaping the thermal performance curve of ectotherms, we predict that the upper limits to performance should covary less with mean temperature conditions than with a measure of variation in temperature, such as diurnal or interannual variation in extremes of temperature. Theory demonstrates that relatively small increases in temperature variance are strong predictors of an increase in the likelihood of extreme events (Katz and Brown 1992; Meehl et al. 2000; Jentsch et al. 2007). Similarly, if high precipitation and cloud cover strongly affect thermoregulatory capability and energy flow relationships, we predict that optimal or preferred body temperatures  $(T_p)$  should be more related to these climate variables than to mean temperature conditions. Moreover, thermoregulatory constrains due to increased rainfall should result in a negative relationship between  $T_{\rm p}$  and rainfall, whereas dehydration and low energy resources in arid areas should result in a positive relationship.

To examine the extent to which temperature variation, precipitation regimes, and their interaction are likely to influence thermal performance curves, we focus on squamate reptiles (mostly lizards and snakes), using a macrophysiological approach (Gaston et al. 2009). In particular, we assess at a global scale how the preferred body temperature ( $T_p$ : the temperature selected by an animal in a laboratory thermal gradient that provides a broad range of thermal opportunities, free of physical and biotic constraints, and that generally reflects the optimal temperature for physiological processes; Dawson 1975; Stevenson et al. 1985; Bauwens et al. 1995), the lower critical temperature for movement (CT<sub>min</sub>: critical thermal minimum), and the upper critical temperature for movement (CT<sub>max</sub>: critical thermal maximum) relate to a set of four climate variable groups and their interactions. The four groups include (1) mean temperature conditions, (2) temperature variation, (3) mean precipitation conditions, and (4) precipitation seasonality. We use an information-theoretic approach (Burnham and Anderson 2002) to identify the most likely model for the relationship between performance parameters and environmental variables given the a priori predictions while also considering potential phylogenetic effects. In addition, we test models that include ecological and life-history variables that are known to affect critical temperature limits and preferred body temperature or that influence the thermal environment experienced by these species and therefore may influence the performance curves.

#### Methods

#### Data Selection

We used data on critical thermal maximum (CT<sub>max</sub>), critical thermal minimum (CT<sub>min</sub>), and preferred body temperature  $(T_p)$  of squamates compiled from the literature (652 data points from 396 species in 16 squamate families; available in Dryad, http://dx.doi.org/10.5061/dryad.8665). Data were not included in the database if the methods used to determine these parameters were not described or if methods differed substantially from standard protocols (Cowles and Bogert 1944; Licht et al. 1966; Lutterschmidt and Hutchison 1997). In cases where more than one study was found for the same species, priority was given to (1) studies that measured  $CT_{max}$ ,  $CT_{min}$ , and  $T_{p}$  (or  $CT_{max}$  and  $CT_{min}$ ) for the same population; (2) field-fresh individuals over those given short-term acclimations and short-termacclimated individuals over those given long-term acclimations; (3) data taken during active phases (i.e., daytime for diurnal and nighttime for nocturnal species) and active seasons; (4) data taken in a fasted state over a fed state; (5) photothermal and thigmothermal gradients for heliothermic and thigmothermic species, respectively; (6) arithmetic mean over median  $T_{p}$  (due to the higher availability of the former); (7) the loss of righting response over the onset of spasms (OS) as the end point of CT<sub>max</sub> and CT<sub>min</sub> (due to the paucity of OS usage across species; Lutterschmidt and Hutchison 1997); and (8) acclimation at 20° or 25°C when only studies with acclimated individuals were available; if data at these temperatures were not available, the arithmetic mean of acclimation groups was used. If data for several populations were reported for the same species and the criteria above did not favor one study over another, an arithmetic mean was calculated for the species. While some variation associated with method is expected among studies, especially where different investigators are involved, we assume, as have many other macrophysiological investigations, that any error will add random variation to the data rather than bias them significantly in one direction or another (Chown et al. 2003; Huev et al. 2009).

Latitudinal and longitudinal coordinates of the populations sampled for  $CT_{max}$ ,  $CT_{min}$ , and  $T_p$  were determined from the geographic locations given in each respective study. When locations were not given (26% of the data) or multiple populations were used (<1% of the data), the center of the species range was determined using species distribution maps (Cogger 1994; Behler and King 1996; Spawls et al. 2004) and occurrence data from museum specimen records (http://data.gbif.org/). The latitudinal and longitudinal coordinates of the populations sampled for  $CT_{max}$ ,  $CT_{min}$ , and  $T_p$  were used to extract 18 climate variables from WorldClim (30" spatial resolution; records from 1950 to 2000; Hiimans et al. 2005), using ArcView 9.2. The broad geographical nature of this study complicates the use of extreme value theory statistics (Gaines and Denny 1993; Katz et al. 2005), which are often a preferred approach for investigating climate extremes and their impacts. Therefore, we relied on the close relationship between climate variance and extremes (Katz and Brown 1992) and used high-resolution interpolated climate surfaces based on monthly means (Hijmans et al. 2005). The following variables were used: (1) mean temperature conditions: annual mean temperature (Ann\_Ta), mean of the maximum temperature of the warmest month of a year (Max Ta), mean of the minimum temperature of the coldest month of a year (Min\_Ta), mean temperature of warmest quarter, mean temperature of coldest quarter, mean temperature of wettest quarter, and mean temperature of driest quarter; (2) temperature variation: mean diurnal temperature range (i.e., mean of the monthly maximum temperature - minimum temperature), mean annual temperature range (i.e., mean of Max\_Ta – Min\_Ta), and temperature seasonality (i.e., standard deviation of Ann\_Ta  $\times$  100), (3) mean precipitation conditions: annual precipitation, precipitation of wettest month, precipitation of driest month, precipitation of wettest quarter, precipitation of driest quarter, precipitation of warmest quarter, and precipitation of coldest quarter; and (4) precipitation variation: precipitation seasonality (coefficient of variation; for additional information see http:// www.worldclim.org).

The relationships between the parameters of the performance curves and environmental variables are likely to be influenced by life-history variation. First, nocturnal species generally have lower  $CT_{max}$  and  $T_p$  than diurnal species (e.g., Greer 1980; Avery 1982; Bennett and John-Alder 1986; Huey et al. 1989), with some exceptions (Huev and Bennett 1987; Kearney and Prevadec 2000). Second, the habitat type used by reptiles strongly affects their capacity to buffer climate changes through behavior (Huey et al. 2003). Therefore, we expect that species most constrained by the availability of optimal microsites will be those most likely to endeavor a shift in performance curves. For example, fossorial or burrowing species should have a lower preferred body temperature than terrestrial or arboreal species (Greer 1980). A phylogenetically explicit analysis that accounts for differences in activity and habitat type across species is needed to assess these effects. Third, reproduction mode (viviparity vs. oviparity) may explain variation in the thermal biology of squamates, as viviparity occurs most frequently in species occupying cold climates, likely providing a selective advantage by maintaining embryos at optimal temperatures via maternal thermoregulation (Shine 2005). Fourth, body mass is associated directly with physiological processes such as digestion and locomotion capacity and indirectly via its influence on the rate of body temperature change (Gates 1980). It is therefore expected that body mass will contribute to variation in thermal performance traits.

The categorical variable "activity" was included in the models to assess the effect of diurnal versus nocturnal activity on performance curve parameters. Crepuscular species were classified as nocturnal species given that in the former, basking is limited. Categories of habitat type were "burrowing" (including fossorial and cryptozoic species), "rock dwelling," "ground dwelling" (mostly terrestrial), "aquatic" (including semiaquatic species), and "arboreal" (including semiarboreal species). The primary habitat type was chosen when populations or species were known to utilize multiple habitat types. Species were also identified as oviparous or viviparous (including ovoviviparous species; Blackburn 2000). For species that have bimodal reproductive status, the mode from the population or region investigated for physiological traits was used in the analyses. Mean body mass (body mass [BM] in g; n = 279 out of a total of 397 species) and snout-vent length (SVL in mm) were taken from studies comprising performance parameters. If these data were not reported, measurements were sought in the literature. In the absence of size data for Masticophis anthonyi, BM of its close relative Masticophis flagellum was used. Allometric equations based on species for which we had both SVL and BM were used to calculate BM for legged-lizard species  $(\log_{10} BM = -4.1877 + 2.7513 \times \log_{10} SVL, R^2 = 0.92,$ n = 108 spp.) and for serpentiform species (e.g., snakes, amphisbaenians, and legless lizard species,  $\log_{10} BM =$  $-3.386 + 1.9293 \times \log_{10}$  SVL,  $R^2 = 0.89$ , n = 30). The use of such equations has some limitations (see Meiri 2010), but inferred body masses represented less than 30% of the data set.

#### Analyses

Body mass and most precipitation variables were  $log_{10}$  transformed given data skewness in preliminary data plots. Ordinary least squares (OLS) and phylogenetic generalized least squares (PGLS) regression methods (Martins and Hansen 1997; Garland and Ives 2000) were then used to examine relationships between the performance curve parameters (i.e.,  $T_p$ ,  $CT_{max}$ , and  $CT_{min}$ ) and the environmental variables. The PGLS analyses adjust for the phylogenetic dependence of the traits investigated by incorporating the expected covariance among species into a model fitted by generalized least squares. PGLS is equivalent to the more

widely used phylogenetic independent contrasts method (Felsenstein 1985; Pagel 1993) when assuming a Brownian motion model of character evolution and using a resolved phylogeny (for a comparison of the two analytical approaches, see Garland and Ives 2000). However, PGLS offers increased flexibility in the specification of the evolutionary model and in the usage of statistical diagnostics typical of standard least squares regressions (Garland and Ives 2000). For example, several alternative evolutionary models can be incorporated by changing the error structure of the models, and both continuous and categorical explanatory variables can be used in the same model (e.g., Halsey et al. 2006). In this study, the covariance matrix was constructed following a Brownian motion model of evolution (Rohlf 2001) and assuming both equal (punctuated model of evolution) and proportional (gradualistic models of evolution) branch lengths in the phylogeny (available in Dryad, http://dx.doi.org/10.5061/dryad .8665). A maximum likelihood approach provided  $\lambda$ , which indicated the degree of phylogenetic correlation in the data ( $\lambda = 0$  indicates no phylogenetic effect, and  $\lambda = 1$  indicates a strong phylogenetic effect equivalent to that expected under the Brownian motion model; for details see Halsey et al. 2006). The best model in the set used to predict a response variable has the lowest Akaike Information Criterion (AIC), and models with  $\Delta$ AIC (model AIC minus that of the best-fit model) <2 are considered as acceptable alternative models; models with  $4 \leq$  $\Delta AIC \leq 7$  have considerable less support, and models with  $\Delta AIC > 10$  have no support (Burnham and Anderson 2001). The Akaike weight,  $w_{p}$ , which is the likelihood of a particular model normalized across the total set of candidate models, was calculated for each model. Akaike weights provide the probability that a model is the best fit among those tested, with values close to 1 being the best models and models with similar weights having similar levels of support in the data (Burnham and Anderson 2001; Johnson and Omland 2004). All the analyses were implemented in R, version 2.11.1 (R Development Core Team 2010), using the APE package (Paradis et al. 2004) and the R function for PGLS from Duncan et al. (2007).

We used several sets of analyses to test for the relationship between performance curve parameters and climate variables. First, we used single-predictor models to test the influence of each climate predictor on each performance variable. We then used low AIC values, low collinearity, and high tolerance levels between variables (Quinn and Keough 2002) to guide variable selection for multiple predictor models. Each model started with the lowest singlepredictor AIC value, and the next strongest predictor was sequentially added. In line with the a priori predictions, we tested for models that included solely variables relating to (1) mean temperature conditions, (2) temperature variation, (3) mean precipitation, and (4) precipitation variation and a series of models that include best predictors across climate variable groups (e.g., mean temperature conditions + precipitation seasonality; mean temperature conditions + temperature variation). Generally, variables within each category (e.g., mean temperature conditions) and a few across climate categories were moderately to highly correlated, reducing substantially the number of models tested (correlation matrices are available in Dryad, http://dx.doi.org/10.5061/dryad.8665). Finally, the same approach was used including categorical variables (activity mode, reproduction mode, habitat type) and body mass in the model. Candidate models obtained from these sets of analyses were compared for each performance parameter using Akaike weights since data sets were consistent across models.

Given several statistical problems associated with the coefficient of determination  $(R^2)$  as a measure of goodness of fit in generalized least squares estimations (Blomquist 1980; Judge et al. 1985; and see Lavin et al. 2008 for a more recent discussion), we report  $R^2$  values only from the best predictive nonphylogenetic (OLS) models.

# Results

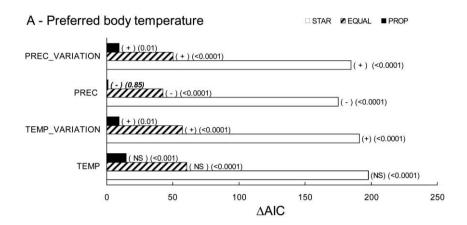
The sampling of squamate species for physiological studies has not been taxonomically uniform. Out of ~7,800 species of squamates globally, we found thermal physiology data for 396 species, of which the Iguanidae and Scincidae included 24% of the species each (193 species in total); the Gekkonidae, Colubridae, Lacertidae, and Agamidae comprised 7%-10% of the species each, and the remaining families (Chamaeleonidae, Varanidae, Viperidae, Teiidae, Cordylidae, Boidae, Xantusidae, Anguidae, Amphisbaenidae, and Helodermatidae) comprised less than 3% of the total number of species each. Similarly, some regions seem to be favored for reptile studies. Generally, Australia, the United States, and southwestern Europe (but not for  $CT_{min}$ ) are well covered compared to the rest of the world (a map with species locations is available in Dryad, http://dx.doi.org/10.5061/dryad.8665).

PGLS models incorporating proportional branch lengths were consistently better than models with no phylogeny (OLS models) or those with equal branch lengths (fig. 1; tables A1, A2 in the online edition of the *American Naturalist*). The consistently high maximum likelihood value  $\lambda$  (>0.85) in PGLS models indicated the high degree of phylogenetic correlation in the data. Precipitation of the driest month was the best single predictor of  $T_p$  in squamate reptiles (fig. 1*A*): areas with low precipitation have species with high  $T_p$  (fig. 2*A*). The inclusion of other climatic variables (mean temperature conditions or temperature variation) barely improved the fit of multivariate

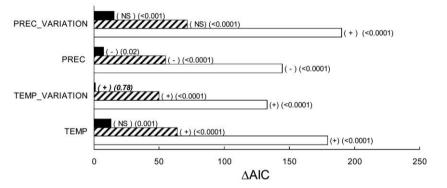
models, and the best model including only mean temperature conditions had a low likelihood as judged by its Akaike weight (<0.0001; table 1; see table A2 for all models tested). The best multivariate predictive model of  $T_{\rm p}$  incorporated precipitation of the driest month (significant negative slope) and body mass (significant negative slope) as continuous variables, a positive interaction between precipitation and body mass and habitat type as a covariate (table 1). Small-sized species tended to have slightly lower  $T_{\rm p}$  than larger species in high-rainfall regions, while larger species had substantially higher  $T_p$  than small species in arid regions (table 2). Negative relationships between  $T_{\rm p}$ and precipitation were consistent across habitat types and while most groups had similar intercepts, the effect of habitat type seemed to originate from the lower  $T_p$  of burrowing species (fig. 2A; table 2). Indeed, a model that groups species from all habitat types (except for burrowing) versus burrowing species improved the model significantly ( $\Delta AIC = 4$ ) and confirmed that burrowing species had on average lower  $T_{\rm p}$  (26.6°C) than nonburrowing species (32.3°C).

Diurnal temperature range was the best single climate predictor of  $CT_{max}$  (fig. 1B): species with high  $CT_{max}$  are located in areas with greater mean diurnal temperature variation (fig. 2B). Variables that described temperature variation were consistently better predictors of CT<sub>max</sub> than were mean temperature variables, and inclusion of mean temperature and/or precipitation variables did not improve the model (table 1; see table A2 for all models tested). The best predictive multivariate model of CT<sub>max</sub> incorporated reproduction mode and habitat type as covariates (table 1). Viviparous species have on average lower  $CT_{max}$  than oviparous species (fig. 2B). Although burrowing species seemed to have lower CT<sub>max</sub> than other habitattype species, a model that clustered burrowing from the rest of habitat-type categories did not improve the model  $(\Delta AIC = 1.4)$ . Instead, the improvement of the initial climate model with inclusion of habitat type as a covariate  $(\Delta AIC = 14)$  suggested that the variation across all habitat types explained a meaningful amount of CT<sub>max</sub> variation.

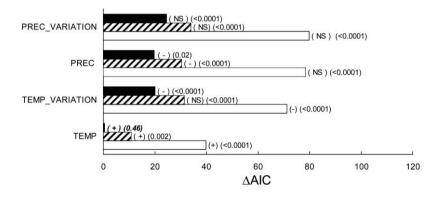
By contrast with  $T_p$  and  $CT_{max}$ , mean temperature conditions were the best single climate predictors of  $CT_{min}$ (fig. 1*C*), indicating that species in areas with warmer average temperature have higher  $CT_{min}$  (fig. 2*C*). Both mean annual temperature and mean temperature of the wettest quarter resulted in similar model outcomes (table A1). The best multivariate model included mean temperature of the wettest quarter (significant positive slope) and temperature seasonality (significant negative slope) as continuous variables and reproduction mode as a categorical variable (table 1; see table A2 for all models tested). Areas with high mean temperature conditions have species with higher  $CT_{min}$  whereas areas with greater temperature sea-



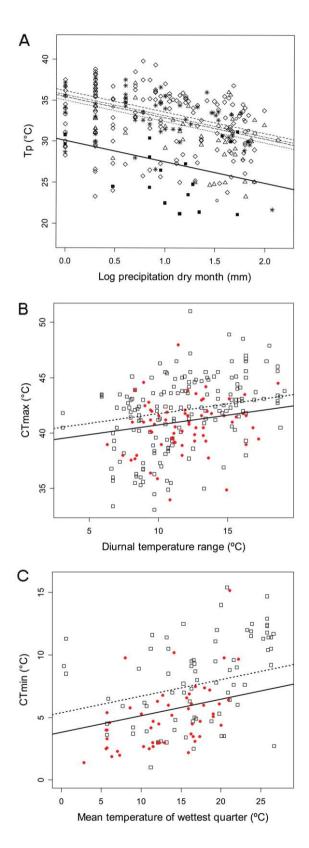
# B - Critical thermal maximum



# C - Critical thermal minimum



**Figure 1:** Rescaled Akaike Information Criterion ( $\Delta$ AIC = model AIC – minimum AIC; Burnham and Anderson 2001) for the best single climate predictor within each climate variable category (precipitation seasonality, mean precipitation conditions, temperature variation, and mean temperature conditions). For each thermal performance curve parameter, the preferred body temperature ( $T_p$ ; A), the critical thermal maximum (CT<sub>max</sub>; B), and the critical thermal minimum (CT<sub>min</sub>; C), three evolutionary models were tested (PROP = proportional branch length, EQUAL = equal branch length phylogeny, and STAR = no phylogeny or ordinary least square analysis). Proportional branch length models were consistently better models, indicated by lower  $\Delta$ AIC. In A,  $T_p$  is best explained by mean precipitation (precipitation of driest month); in B, CT<sub>max</sub> is best explained by temperature variation (diurnal temperature range); and in C, CT<sub>min</sub> is best explained by mean temperature conditions (mean temperature of wettest quarter). For a breakdown and summary of all predictors tested within each climate group, see table A1 in the online edition of the *American Naturalist*. Slope sign and Akaike weights ( $w_i$ ) are shown in parentheses, with the best model italicized.



sonality have species with lower  $CT_{min}$  (table 2). Finally, viviparous species have on average lower  $CT_{min}$  (5°C) than oviparous species (8.1°C; fig. 2*C*).

For  $T_{\rm p}$ , the coefficient of determination ( $R^2$ ) of the best predictive OLS model ( $T_{\rm p} \sim$  precipitation of driest month × body mass + habitat type) was 0.24. When adding taxonomic family as a predictor, the  $R^2$ increased to 0.60. For CT<sub>max</sub>, the  $R^2$  of the best OLS model (CT<sub>max</sub> ~ diurnal temperature range + habitat type + reproduction mode) was 0.26 and 0.45 with Family included. Finally, for CT<sub>min</sub>, the  $R^2$  for the best predictive OLS model (CT<sub>min</sub> ~ mean temperature of wettest quarter + temperature seasonality + reproduction mode) was 0.35, and 0.51 with Family included.

# Discussion

The macrophysiological approach adopted here provides fresh insights into the factors influencing the evolution of performance curves in squamate reptiles. In particular, it is clear that precipitation and temperature variation rather than mean temperature conditions are the best predictors of  $T_p$  and  $CT_{max}$ , respectively, across the group of species investigated. These climate variables have therefore likely played a significant role in the evolution of thermal sensitivity of ectotherm performance.

In general, the thermal preference of an ectotherm closely matches the optimal temperature for performance and thus is the temperature that the organism seeks to achieve, through thermoregulation, to maximize fitness (Angilletta 2009). The influence of precipitation on  $T_p$ indicates that species in habitats with high rainfall have fewer opportunities for behavioral thermoregulation than those in drier habitats. A dearth of optimal microsites for thermoregulation may be found in both temperate and tropical areas. In high-rainfall temperate areas such as the Pacific Coast of the United States, frequent overcast conditions and high rainfall constrain some species to use suboptimal microhabitats (Kingsbury 1994), while the

**Figure 2.** Phylogenetically adjusted scatter plots and analyses indicate that species in locations with high precipitation have a lower preferred body temperature  $(T_p)$  than those in dry areas (*A*), species from areas with high diurnal temperature variation have higher critical thermal maximum (CT<sub>max</sub>) than those in less variable areas (*B*), and species inhabiting areas with high mean temperature of the wettest quarter have higher critical thermal minimum (CT<sub>min</sub>) than those from colder habitats (*C*). *A*, Burrowing species (*filled squares*; *solid line*) have lower  $T_p$  than other habitat-type dwellers (aquatic [*circle*], arboreal [*triangle*], ground- [*diamond*] and rock-dwellers [*asterisk*]; all represented by thin lines). Viviparous species (*red circles*; *solid line*) have lower CT<sub>max</sub> (*B*) and lower CT<sub>min</sub> (*C*) than oviparous species (*open squares*; *stippled line*).

iables for each performance curve parameter				
Model	λ	AIC	$W_i$	
Preferred temperature $(T_p)$ :				
1. Precipitation driest month $\times$ mass + habitat	.89	1,327.38	.79	
2. Mean annual temperature + mass + habitat	.89	1,349.28	<.0001	
3. Mass + habitat	.89	1,348.53	<.0001	
Critical thermal maximum (CT <sub>max</sub> ):				
1. Diurnal temperature range + habitat +				
reproduction	.93	1,012.73	.70	
2. Mean annual temperature + habitat +				
reproduction	.94	1,029.44	<.001	
3. Habitat + reproduction	.94	1,028.48	<.001	
Critical thermal minimum (CT <sub>min</sub> ):				
1. Mean temperature of the wettest quarter + tem-				
perature seasonality + reproduction	.94	601.74	.56	
2. Precipitation coldest quarter + reproduction	.87	624.99	<.0001	
3. Reproduction	.87	625.56	<.0001	

 Table 1: Best predictive models incorporating climate variables and life-history variables for each performance curve parameter

Note: For each parameter, we present the best model overall (1), the best model including mean temperature conditions (or precipitation in the case of  $CT_{min}$  since temperature was the best climate predictor for this parameter; 2), and the best model including life-history variables only (3). All models include a proportional phylogeny branch-length assumption. The categorical variable habitat distinguishes across species habitat-type usage, and reproduction indicates reproductive mode. Body mass (mass) and precipitation variables are log transformed.  $\lambda =$  phylogenetic dependence; AIC = Akaike Information Criterion; and  $w_i$  = Akaike weights calculated from AIC values of all tested models given in table A2 in the online edition of the American Naturalist.

presence of dense-canopy forests in tropical areas constrain reptiles to use shaded microsites despite the potential availability of gaps and edges (Hertz 1974; Huey 1974). Given the high dependence of heat-transfer budgets of ectotherms on their surrounding physical environment (Gates 1980), the reduced availability of optimal operative temperatures in areas with high rainfall restricts activity periods and likely imposes lower body temperatures (e.g., Marquis et al. 2008; Clusella-Trullas et al. 2009). Similarly, the reduction in available operative temperatures favorable for activity in arid areas may force individuals to select high body temperatures in these regions (e.g., Grant 1990). Moreover, the similarity of effects of precipitation on  $T_{\rm p}$ for burrowing species suggests that the mechanism underlying this relationship may involve more than the reduced availability of basking sites (i.e., direct solar radiation). Instead, fossorial species in high rainfall areas are likely to be as much constrained by suboptimal microsites as are species that spend more time above ground. Despite the relationship of body mass and  $T_{\rm p}$  being variable across rainfall gradients (i.e., a body mass  $\times$  T<sub>p</sub> interaction), possibly due to the mixed effects that water availability can have on growth rate, adult body size, fecundity, and predation rates (e.g., Tinkle et al. 1993; Sears and Angilletta 2003, 2004), precipitation consistently remained the best predictor of  $T_p$ , reinforcing its marked influence on thermoregulation opportunities.

For CT<sub>max</sub>, variables that reflect temperature variation were consistently better predictors than mean temperature variables. Because an increase in temperature variance is also a strong indicator of temperature extremes (Katz and Brown 1992; Meehl et al. 2000), this might reflect selection for survival of occasional extremes rather than simply variation in temperature range, especially because diurnal temperature range is a consequence of variation in both minimum and maximum temperature, of which the former tends to be most pronounced (Bonan 2002; Chown et al. 2004). That strong selection for increasing CT<sub>max</sub> might be a consequence of extreme events is in keeping with theory (see "Introduction") and indicates that some variation for change in this trait might exist despite its known limited range compared to lower critical limits (see discussion in Chown 2001; Hoffmann et al. 2003). Moreover the direction of these climate effects is consistent across habitat-type groups and between reproductive modes. Although viviparity is thought to have evolved in cold and/or unpredictable climates to enhance fitness (e.g., cold climate hypothesis, maternal manipulation hypothesis; Tinkle and Gibbons 1977; Shine 1995, 2004; but see Crespi and Semeniuk 2004), strong selection for increased CT<sub>max</sub> (and lower CT<sub>min</sub>) in more variable environments is likely in viviparous species.

Predictions of animal, and especially ectotherm, responses to ongoing and forecast climate change have

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 Table 2: Predictor coefficients for the best-fit models for each

 performance curve parameter

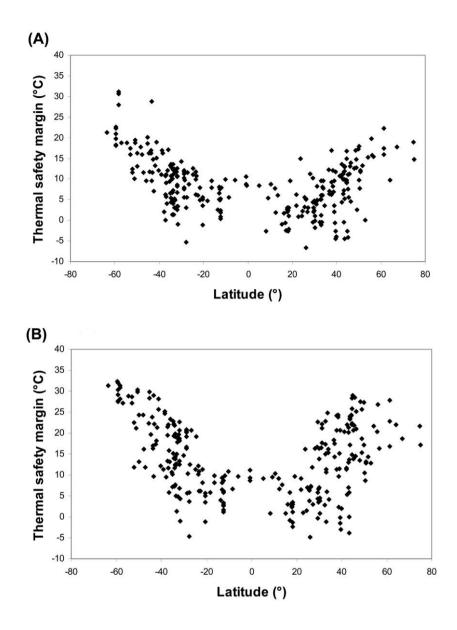
	Estimate	SE
Preferred temperature:		
Log precipitation driest month	-2.59	.60
Log mass	-1.62	.49
Log precipitation driest month × log		
mass	.81	.33
Habitat:		
Aquatic species	36.16	1.93
Arboreal species	35.01	.98
Burrowing species	30.11	1.31
Ground-dwelling species	35.71	.88
Rock-dwelling species	35.45	.93
Critical thermal maximum:		
Diurnal temperature range	.18	.04
Habitat:		
Aquatic species	42.01	1.65
Arboreal species	40.76	.95
Burrowing species	37.84	1.07
Ground-dwelling species	40.16	.88
Rock-dwelling species	40.84	1.01
Reproduction:		
Oviparous	40.01	1.43
Viviparous	38.99	.42
Critical thermal minimum:		
Mean temperature of the wettest quarter	.14	.03
Temperature seasonality	003	.001
Reproduction:		
Oviparous	6.67	1.65
Viviparous	5.10	.51

placed much emphasis on temperature change and especially on increases in mean temperature (e.g., Thomas et al. 2004; Parmesan 2006; Musolin 2007; Huey et al. 2009). However, a major implication of our results is that forecasts of the effects of future climate change on terrestrial reptiles must take changes in precipitation and temperature variation into account. For example, if the evolved variation in  $T_p$  is largely a consequence of variation in precipitation, and if mean temperature increases in tropical regions are accompanied by increasing precipitation and/or cloud cover (Solomon et al. 2007; Adler et al. 2008), changes in mean thermal conditions may have little influence on the thermal safety margin of ectotherms in these regions, unlike previous forecasts (Deutsch et al. 2008; Huey et al. 2009). By contrast, forecast reductions in midlatitude precipitation and cloud area fraction (see "Introduction"; Fung et al. 2011) suggest that species in these areas will be more at risk than previously thought. Indeed, if a thermal safety margin plot of the kind previously used to indicate risks to tropical ectotherm species (Deutsch et al. 2008; i.e.,  $T_{\rm p}$  – ambient temperature) is constructed using mean temperature of the warmest or wettest quarter

(the times of greatest reptile activity; Pianka and Vitt 2006) or using mean annual temperature (as in Deutsch et al. 2008), it is clear that species in midlatitudes, rather than in the tropics, are those already most at risk (fig. 3).

Further complexity is introduced when predicting impacts of climate change on performance curves by the fact that changes in diurnal temperature range, the primary climatic correlate of squamate  $CT_{max}$ , show strong spatial structure. While a decreasing diurnal temperature range is largely accompanied by increased cloud cover and precipitation in terrestrial areas, the most pronounced reductions in diurnal temperature range are taking place over arid regions or those with sparse or short vegetation, largely as a consequence of increases in daily minimum temperature. Here, greater than average warming and reductions in precipitation and cloud cover are also expected (Vose et al. 2005; Zhou et al. 2009a). Therefore, in these regions, one climatic signature that should promote further selection for resistance to unpredictable climate extremes in species already at the margins of their tolerance (fig. 3) is weakening. Given the positive relationship between  $T_p$  and  $CT_{max}$  ( $R^2 = 0.55$ , P < .0001; plot available in Dryad, http://dx.doi.org/10.5061/dryad.8665), these environmental influences on CT<sub>max</sub> are likely to interact with those affecting  $T_p$ . However, relatively high variation around the relationship between  $CT_{max}$  and  $T_p$  and differences in the likelihood of trait evolution in ectotherm generalists versus specialists (Kellerman et al. 2009) imply that considerable complexity of future evolved responses to climate change is likely. Although frequently neglected in modeling approaches, evolutionary dynamics are likely to be critical for species responses to changing climates (Kellerman et al. 2009).

As opposed to the two other parameters of the performance curve, both mean temperature conditions and temperature variation were the strongest predictors of lower temperature limits. These findings are closest to what intuitively might be expected: lower limits to ectotherm activity are likely to be set by ambient temperatures because initial movement for thermoregulation is dependent on ambient temperatures at the start of activity (Cowles and Bogert 1944) and by temperature variation because higher frequency or magnitude of temperature extremes may impose additional selection on the lower margin of performance (Levins 1968; Watson and Hoffmann 1996). Increasing ambient temperatures in future climate scenarios would thus seem generally to have little negative significance for the lower limits to ectotherm performance. However, in temperate areas where increases in mean ambient temperature are not being accompanied by changes in the timing of low temperature extremes in the spring (e.g., Mulholland et al. 2009), any selection for an increase in CT<sub>min</sub> as a consequence of increasing ambient temperatures



**Figure 3:** Plot of thermal safety margin (following Deutsch et al. 2008) against latitudinal position adjusted for altitude. *A*, Thermal safety margin is calculated as the difference between optimal temperature (here, preferred temperature  $T_p$ ) and mean temperature of the warmest quarter. Species at midlatitudes have the lowest safety margin, indicating that the optimum body temperature for these species is closest to average ambient temperature conditions, increasing the likelihood of reduced performance with warming. *B*, This pattern is consistent when thermal safety margin is calculated with mean annual temperature.

may in fact have considerable negative impacts on performance.

In conclusion, the results of this study demonstrate that variation in environmental temperatures and variation in precipitation have had significant influences on two major components of reptile thermal performance curves. The influence of mean conditions on the third major variable of thermal performance curves, CT<sub>min</sub>, may also help account for complex relationships among these performance

traits (Angilletta 2009; figures with these relationships are available in Dryad, http://dx.doi.org/10.5061/dryad.8665). Overall, these results show that the impacts of climate change on ectotherms will not be straightforward to predict. Spatially variable forecasts for changes in climate variables, the local effects of humans on these variables (Nair et al. 2003), and habitat structure (Zhou et al. 2009*b*), coupled with variation in the evolutionary responses of species (Kellermann et al. 2009), make this especially likely.

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Recent smaller-scale studies of other groups (Helmuth et al. 2006; Mulholland et al. 2009; Zimmermann et al. 2009), have also demonstrated that biological responses to environmental change are not as straightforward as some earlier considerations might have suggested. Thus, current uncertainty regarding the biological consequences of climate change is perhaps larger than generally acknowledged, emphasising that strategies for mitigation of climate change may be more effective for long-term conservation than those for assisting biodiversity to adapt, either through relocation or other interventions.

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## Literature Cited

- Addo-Bediako, A., S. L. Chown, and K. J. Gaston. 2000. Thermal tolerance, climatic variability and latitude. Proceedings of the Royal Society B: Biological Sciences 267:739–745.
- Adler, R. F., G. Gu, J.-J. Wang, G. J. Huffman, S. Curtis, and D. Bolvin. 2008. Relationships between global precipitation and surface temperature on interannual and longer timescales (1979– 2006). Journal of Geophysical Research 113:D22104, doi:10.1029/ 2008JD010536.
- Angilletta, M. J., Jr. 2009. Thermal adaptation: a theoretical and empirical synthesis. Oxford University Press, Oxford.
- Avery, R. A. 1971. Estimates of food consumption by the lizard Lacerta vivipara. Journal of Animal Ecology 40:351–365.
- . 1982. Field studies of body temperature and thermoregulation. Pages 93–166 *in* C. Gans and F. H. Pough, eds. Biology of the reptilia. Vol. 12. Physiological ecology. Academic Press, London.
- Bauwens, D., T. Garland Jr., A. M. Castilla, and R. Van Damme. 1995. Evolution of sprint speed in lacertid lizards: morphological, physiological, and behavioural covariation. Evolution 49:848–863.
- Behler, J. L., and F. W. King. 1996. National Audubon Society field guide to North American reptiles and amphibians. Knopf, New York.
- Bennett, A. F., and H. John-Alder. 1986. Thermal relations of some Australian skinks. Copeia 1986:57–64.
- Blackburn, D. G. 2000. Reptilian viviparity: past research, future directions, and appropriate models. Comparative Biochemistry and Physiology A 127:391–409.
- Blomquist, N. S. 1980. A note on the use of the coefficient of determination. Scandinavian Journal of Economics 82:409–412.
- Blouin-Demers, G., and P. J. Weatherhead. 2001. An experimental test of the link between foraging, habitat selection and themoregulation in black rat snakes *Elaphe obsoleta obsoleta*. Journal of Animal Ecology 70:1006–1013.

- Bonan, G. 2002. Ecological climatology: concepts and applications. Cambridge University Press, Cambridge.
- Bonebrake, T. C., and M. D. Mastrandrea. 2010. Tolerance adaptation and precipitation changes complicate latitudinal patterns of climate change impacts. Proceedings of the National Academy of Sciences of the USA 107:12581–12586.
- Bradshaw, W. E., and C. M. Holzapfel. 2006. Evolutionary response to rapid climate change. Science 312:1477–1478.
- Brown, C. R., and M. B. Brown. 1998. Intense natural selection on body size and wing and tail asymmetry in cliff swallows during severe weather. Evolution 52:1461–1475.
- Brown, R. P., and S. Griffin. 2005. Lower selected body temperatures after food deprivation in the lizard *Anolis carolinensis*. Journal of Thermal Biology 30:79–83.
- Burnham, K. P., and D. R. Anderson. 2001. Kullback-Leiber information as a basis for strong inference in ecological studies. Wildlife Research 28:111–119.
- Chen, I. C., H. Shiu, S. Benedick, J. D. Holloway, V. K. Chey, H. S. Barlow, J. K. Hill, and C. D. Thomas. 2009. Elevation increases in moth assemblages over 42 years on a tropical mountain. Proceedings of the National Academy of Sciences of the USA 106: 1479–1483.
- Chown, S. L. 2001. Physiological variation in insects: hierarchical levels and implications. Journal of Insect Physiology 47:649–660.
- Chown, S. L., A. Addo-Bediako, and K. J. Gaston. 2003. Physiological diversity: listening to the large-scale signal. Functional Ecology 17: 562–572.
- Chown, S. L., B. J. Sinclair, H. P. Leinaas, and K. J. Gaston. 2004. Hemispheric asymmetries in biodiversity: a serious matter for ecology. PLoS Biology 2:1701–1707.
- Clusella-Trullas, S., J. H. van Wyk, and J. R. Spotila. 2009. Thermal benefits of melanism in cordylid lizards: a theoretical and field test. Ecology 90:2297–2312.
- Cogger, H. G. 1994. Reptiles and amphibians of Australia. Cornell University Press, Ithaca, NY.
- Cowles, R. B., and C. M. Bogert. 1944. A preliminary study of the thermal requirements of desert reptiles. Bulletin of the American Museum of Natural History 83:265–296.
- Crespi, B., and C. Semeniuk. 2004. Parent-offspring conflict in the evolution of vertebrate reproductive mode. American Naturalist 163:635–653.
- Dawson, W. R. 1975. On the physiological significance of the preferred body temperatures of reptiles. Pages 443–473 *in* D. M. Gates and R. B. Schmerl, eds. Perspectives in biophysical ecology. Springer, Berlin.
- Deutsch, C. A., J. J. Tewksbury, R. B. Huey, K. S. Sheldon, C. K. Ghalambor, D. C. Haak, and P. R. Martin. 2008. Impacts of climate warming on terrestrial ectotherms across latitude. Proceedings of the National Academy of Sciences of the USA 105:6668–6672.
- Duncan, R. P., D. M. Forsyth, and J. Hone. 2007. Testing the metabolic theory of ecology: allometric scaling exponents in mammals. Ecology 88:324–333.
- Dunham, A. E. 1978. Food availability as a proximate factor influencing individual growth rates in the iguanid lizard *Sceloporus merriami*. Ecology 59:770–778.
- Endler, J. A. 1986. Natural selection in the wild. Princeton University Press, Princeton, NJ.
- Felsenstein, J. 1985. Phylogenies and the comparative method. American Naturalist 125:1–15.
- Fischer, D. T., C. J. Still, and A. P. Williams. 2009. Significance of summer fog and overcast for drought stress and ecological func-

tioning of coastal California endemic plant species. Journal of Biogeography 36:783–799.

- Forcada, J., P. N. Trathan, K. Reid, E. J. Murphy, and J. P. Croxall. 2006. Contrasting population changes in sympatric penguin species in association with climate warming. Global Change Biology 12:411–423.
- Fung, F., A. Lopez, and M. New. 2011. Water availability in +2°C and +4°C worlds. Philosophical Transactions of the Royal Society A 369:99–116.
- Gaines, S. D., and M. W. Denny. 1993. The largest, smallest, highest, lowest, longest, and shortest: extremes in ecology. Ecology 74: 1677–1692.
- Garland, T., Jr., and A. R. Ives. 2000. Using the past to predict the present: confidence intervals for regression equations in phylogenetic comparative methods. American Naturalist 155:346–364.
- Gaston, K. J., S. L. Chown, P. Calosi, J. Bernardo, D. T. Bilton, A. Clarke, S. Clusella-Trullas, et al. 2009. Macrophysiology: a conceptual reunification. American Naturalist 174:595–612.
- Gates, D. M. 1980. Biophysical ecology. Springer, New York.
- Ghalambor, C. K., R. B. Huey, P. R. Martin, J. J. Tewksbury, and G. Wang. 2006. Are mountain passes higher in the tropics? Janzen's hypothesis revisited. Integrative and Comparative Biology 46:5–17.
- Gilchrist, G. W. 1995. Specialists versus generalists in changing environments. I. Fitness landscapes of thermal sensitivity. American Naturalist 146:252–270.
- Grant, B. R., and P. R. Grant. 1993. Evolution of Darwin's finches caused by a rare climatic event. Proceedings of the Royal Society B: Biological Sciences 251:111–117.
- Grant, B. W. 1990. Trade-offs in activity time and physiological performance for thermoregulating desert lizards, *Sceloporus merriami*. Ecology 71:2323–2333.
- Greer, A. E. 1980. Critical thermal maximum temperatures in Australian scincid lizards: their ecological and evolutionary significance. Australian Journal of Zoology 28:91–102.
- Gutschick, V. P., and H. BassiriRad. 2003. Extreme events as shaping physiology, ecology, and evolution of plants: towards a unified definition and evaluation of their consequences. New Phytologist 160:21–42.
- Halsey, L. G., P. J. Butler, and T. M. Blackburn. 2006. A phylogenetic analysis of the allometry of diving. American Naturalist 167:276– 287.
- Harlow, H. J., S. S. Hillman, and M. Hoffman. 1976. The effect of temperature on the digestive efficiency in the herbivorous lizard *Dipsosaurus dorsalis*. Journal of Comparative Physiology B 111:1–6.
- Helmuth, B., N. Mieszkowska, P. Moore, and S. J. Hawkins. 2006. Living on the edge of two changing worlds: forecasting the responses of rocky intertidal ecosystems to climate change. Annual Review of Ecology, Evolution, and Systematics 37:373–404.
- Hertz, P. E. 1974. Thermal passivity of a tropical forest lizard, *Anolis polylepis*. Journal of Herpetology 8:323–327.
- . 1992. Temperature regulation in Puerto Rican *Anolis* lizards: a field test using null hypotheses. Ecology 73:1405–1417.
- Hijmans, R. J., S. E. Cameron, J. L. Parra, P. G. Jones, and A. Jarvis. 2005. Very high resolution interpolated climate surfaces for global land areas. International Journal of Climatology 25:1965–1978.
- Hoegh-Guldberg, O., L. Hughes, S. McIntyre, D. B. Lindenmayer, C. Parmesan, H. P. Possingham, and C. D. Thomas. 2008. Assisted colonization and rapid climate change. Science 321:345–346.
- Hoffmann, A. A., and P. A. Parsons. 1997. Extreme environmental change and evolution. Cambridge University Press, Cambridge.
- Hoffmann, A. A., J. G. Sorensen, and V. Loeschcke. 2003. Adaptation

of *Drosophila* to temperature extremes: bringing together quantitative and molecular approaches. Journal of Thermal Biology 213: 870–880.

- Huey, R. B. 1974. Behavioral thermoregulation in lizards: importance of associated costs. Science 184:1001–1003.
- Huey, R. B., and A. F. Bennett. 1987. Phylogenetic studies of coadaptation: preferred temperatures versus optimal performance temperatures in ectotherms. Evolution 41:109–115.
- Huey, R. B., P. H. Niewiarowski, J. Kaufmann, and J. C. Herron. 1989. Thermal biology of nocturnal ectotherms: is sprint performance of geckos maximal at low body temperatures? Physiological Zoology 62:488–504.
- Huey, R. B., P. E. Hertz, and B. Sinervo. 2003. Behavioral drive versus behavioral inertia in evolution: a null model approach. American Naturalist 161:357–366.
- Huey, R. B., C. A. Deutsch, J. J. Tewksbury, L. J. Vitt, P. E. Hertz, H. J. Alvarez Pérez, and T. Garland Jr. 2009. Why tropical forest lizards are vulnerable to climate warming. Proceedings of the Royal Society B: Biological Sciences 276:1939–1948.
- Janzen, D. J. 1967. Why mountain passes are higher in the tropics. American Naturalist 101:203–249.
- Jentsch, A., J. Kreyling, and C. Beierkuhnlein. 2007. A new generation of climate-change experiments: events, not trends. Frontiers in Ecology and the Environment 5:365–374.
- Johnson, J. B., and K. S. Omland. 2004. Model selection in ecology and evolution. Trends in Ecology & Evolution 19:101–108.
- Judge, G. G., W. E. Griffiths, R. C. Hill, H. Lutkepohl, and T.-C. Lee. 1985. The theory and practice of econometrics. Wiley, New York.
- Katz, R. W., and B. G. Brown. 1992. Extreme events in a changing climate: variability is more important than averages. Climatic Change 21:289–302.
- Katz, R. W., G. S. Brush, and M. B. Parlange. 2005. Statistics of extremes: modelling ecological disturbances. Ecology 86:1124–1134.
- Kearney, M., and M. Prevadec. 2000. Do nocturnal ectotherms thermoregulate? a study of the temperate gecko *Christinus marmoratus*. Ecology 81:2984–2996.
- Kearney, M., R. Shine, and W. P. Porter. 2009. The potential for behavioural thermoregulation to buffer "cold-blooded" animals against climate warming. Proceedings of the National Academy of Sciences of the USA 106:3835–3840.
- Kellermann, V., B. van Heerwaarden, C. Sgrò, and A. A. Hoffmann. 2009. Fundamental evolutionary limits in ecological traits drive *Drosophila* species distributions. Science 325:1244–1246.
- Kingsbury, B. A. 1994. Thermal constraints and eurythermy in the lizard *Elgaria multicarinata*. Herpetologica 50:266–273.
- Knapp, A. K., C. Beier, D. D. Briske, A. T. Classen, Y. Luo, M. Reichstein, M. D. Smith, et al. 2008. Consequences of more extreme precipitation regimes for terrestrial ecosystems. BioScience 58:811–821.
- Lavin, S. R., W. H. Karasov, A. R. Ives, K. M. Middleton, and T. Garland Jr. 2008. Morphometrics of the avian small intestine compared with that of nonflying mammals: a phylogenetic approach. Physiological and Biochemical Zoology 81:526–550.
- Levins, R. 1968. Evolution in changing environments. Princeton University Press, Princeton, NJ.
- Licht, P., W. R. Dawson, V. H. Shoemaker, and A. R. Main. 1966. Observations on the thermal relations of western Australian lizards. Copeia 1966:97–110.
- Lorenzon, P., J. Clobert, A. Oppliger, and H. John-Alder. 1999. Effect of water constraint on growth rate, activity and body temperature

of yearling common lizard (*Lacerta vivipara*). Oecologia (Berlin) 118:423–430.

- Lorenzon, P., J. Clobert, and M. Massot. 2001. The contribution of phenotypic plasticity to adaptation in *Lacerta vivipara*. Evolution 55:392–404.
- Lutterschmidt, W. I., and V. H. Hutchison. 1997. The critical thermal maximum: data to support the onset of spasms as the definitive end point. Canadian Journal of Zoology 75:1553–1560.
- Lynch, M., and W. Gabriel. 1987. Environmental tolerance. American Naturalist 129:283–303.
- Marquis, O., M. Massot, and J. F. Le Galliard. 2008. Intergenerational effects of climate generate cohort variation in lizard reproductive performance. Ecology 89:2575–2583.
- Martins, E. P., and T. F. Hansen. 1997. Phylogenies and the comparative method: a general approach to incorporating phylogenetic information into the analysis of interspecific data. American Naturalist 144:193–209.
- McKechnie, A. E., and B. O. Wolf. 2010. Climate change increases the likelihood of catastrophic avian mortality events during extreme heat waves. Biology Letters 6:253–256.
- Meehl, G. A., T. Karl, D. R. Easterling, S. Changnon, R. Pielke Jr., D. Changnon, J. Evans, et al. 2000. An introduction to trends in extreme weather and climate events: observations, socioeconomic impacts, terrestrial ecological impacts, and model projections. Bulletin of the American Meteorological Society 81:413–416.
- Meehl, G. A., T. F. Stocker, W. D. Collins, P. Friedlingstein, A. T. Gaye, J. M. Gregory, A. Kitoh, et al. 2007. Global climate projections. Pages 747–845 *in* Climate change 2007: the physical science basis. Contribution of Working Group I to the fourth assessment report of the Intergovernmental Panel on Climate Change. S. Solomon, D. Qin, M. Manning, Z. Chen, M. Marquis, K. B. Averyt, M. Tignor, and H. L. Miller, eds. Cambridge University Press, Cambridge.
- Meiri, S. 2010. Length-weight allometries in lizards. Journal of Zoology 281:218–226.
- Mulholland, P. J., B. J. Roberts, W. R. Hill, and J. G. Smith. 2009. Stream ecosystem responses to the 2007 spring freeze in the southeastern United States: unexpected effects of climate change. Global Change Biology 15:1767–1776.
- Musolin, D. L. 2007. Insects in a warmer world: ecological, physiological and life-history responses of true bugs (Heteroptera) to climate change. Global Change Biology 13:1565–1585.
- Nair, U. S., R. O. Lawton, R. M. Welch, and R. A. Pielke Sr. 2003. Impact of land use on Costa Rican tropical montane cloud forests: sensitivity of cumulus cloud field characteristics to lowland deforestation. Journal of Geophysical Research 108:D74206, doi: 10.1029/2001JD001135.
- Pagel, M. D. 1993. Seeking the evolutionary regression coefficient: an analysis of what comparative methods measure. Journal of Theoretical Biology 164:191–205.
- ———. 1999. Inferring the historical patterns of biological evolution. Nature 401:877–884.
- Paradis, E., J. Claude, and K. Strimmer, 2004. APE: analyses of phylogenetics and evolution in R language. Bioinformatics 20:289–290.
- Parker, B. R., R. D. Vinebrooke, and D. W. Schindler. 2008. Recent climate extremes alter alpine lake ecosystems. Proceedings of the National Academy of Sciences of the USA 105:12927–12931.
- Parmesan, C. 2006. Ecological and evolutionary responses to recent climate change. Annual Review of Ecology, Evolution, and Systematics 37:637–669.
- Parmesan, C., T. L. Root, and M. R. Willig. 2000. Impacts of extreme

weather and climate on terrestrial biota. Bulletin of the American Meteorological Society 81:443–450.

- Pelini, S. L., J. D. K. Dzurisin, K. M. Prior, C. M. Willams, T. D. Marisco, B. J. Sinclair, and J. J. Hellmann. 2009. Translocation experiments with butterflies reveal limits to enhancement of poleward populations under climate change. Proceedings of the National Academy of Sciences of the USA 106:11160–11165.
- Pianka, E. R., and L. J. Vitt. 2006. Lizards: windows to the evolution of diversity. University of California Press, Berkeley.
- Pounds, J. A., M. R. Bustamante, L. A. Coloma, J. A. Consuegra, M. P. Fogden, P. N. Foster, E. La Marca, et al. 2006. Widespread amphibian extinctions from epidemic disease driven by global warming. Nature 439:161–167.
- Quinn, G. P., and M. J. Keough. 2002. Experimental design and data analysis for biologists. Cambridge University Press, Cambridge.
- R Development Core Team. 2010. R: a language and environment for statistical computing. R Foundation for Statistical Computing, Vienna, http://www.R-project.org.
- Raxworthy, C. J., R. G. Pearson, N. Rabibisoa, A. M. Rakotondrazafy, J.-B. Ramanamanjato, A. P. Raselimanana, S. Wu, R. A. Nussbaum, and D. A. Stone. 2008. Extinction vulnerability of tropical montane endemism from warming and upslope displacement: a preliminary appraisal for the highest massif in Madagascar. Global Change Biology 14:1703–1720.
- Rohlf, F. J. 2001. Comparative methods for the analysis of continuous variables: geometric interpretations. Evolution 55:2143–2160.
- Rovito, S. M., S. Parra-Olea, C. R. Vásquez-Almazán, T. J. Papenfuss, and D. B. Wake. 2009. Dramatic declines in Neotropical salamander populations are an important part of the global amphibian crisis. Proceedings of the National Academy of Sciences of the USA 106:3231–3236.
- Sears, M. W., and M. J. Angilletta Jr. 2003. Life-history variation in the sagebrush lizard: phenotypic plasticity or local adaptation? Ecology 84:1624–1634.
- 2004. Body size clines in *Sceloporus* lizards: proximate mechanisms and demographic constraints. Integrative and Comparative Biology 44:433–442.
- Shine, R. 1995. A new hypothesis for the evolution of viviparity in reptiles. American Naturalist 160:582–593.
- 2004. Does viviparity evolve in cold climate reptiles because pregnant females maintain stable (not high) body temperatures? Evolution 58:1809–1818.
- 2005. Life-history evolution in reptiles. Annual Review of Ecology, Evolution, and Systematics 36:23–46.
- Sinervo, B., F. Mendez-de-la-Cruz, D. B. Miles, B. Heulin, E. Bastiaans, M. Villagran-Santa Cruz, R. Lara-Resendiz, et al. 2010. Erosion of lizard diversity by climate change and altered thermal niches. Science 328:894–899.
- Snyder, G. K., and W. W. Weathers. 1975. Temperature adaptations in amphibians. American Naturalist 109:93–101.
- Solomon, S. D., D. Qin, M. Manning, Z. Chen, M. Marquis, K. B. Averyt, M. Tignor, and H. L. Miller. 2007. Climate change 2007: the physical science basis. Contribution of Working Group I to the fourth assessment report of the Intergovernmental Panel on Climate Change. Cambridge University Press, Cambridge.
- Spawls, S., K. Howell, C. Drewes, and J. Ashe. 2004. A field guide to the reptiles of east Africa. C. Helm, London.
- Stachowicz, J. J., J. R. Terwin, R. B. Whitlatch, and R. W. Osman. 2002. Linking climate change and biological invasions: ocean warming facilitates nonindigenous species invasions. Proceedings of the National Academy of Sciences of the USA 99:15497–15500.

- Stevens, G. C. 1989. The latitudinal gradient in geographical range: how so many species coexist in the tropics. American Naturalist 133:240–256.
- Stevenson, R. D., C. R. Peterson, and J. T. Tsuji. 1985. The thermal dependence of locomotion, tongue flicking, digestion and oxygen consumption in the wandering garter snake. Physiological Zoology 58:46–57.
- Thomas, C. D., A. Cameron, R. E. Green, M. Bakkenes, L. J. Beaumont, Y. C. Collingham, B. F. N. Erasmus, et al. 2004. Extinction risk from climate change. Nature 427:145–148.
- Tinkle, D. W., and J. M. Gibbons. 1977. The distribution and evolution of viviparity in reptiles. Miscellaneous Publications, Museum of Zoology, University of Michigan 154:1–55.
- Tinkle, D. W., A. E. Dunham, and J. D. Congdon. 1993. Life history and demographic variation in the lizard *Sceloporus graciosus*: a long-term study. Ecology 74:2413–2429.
- Trenberth, K. E., L. Smith, T. Qian, A. Dai, and J. Fasullo. 2007. Estimates of the global water budget and its annual cycle using observational and model data. Journal of Hydrometeorology 8: 758–768.
- Vose, R. S., D. R. Easterling, and B. Gleason. 2005. Maximum and minimum temperature trends for the globe: an update through 2004. Geophysical Research Letters 32:L23822, doi:10.1029/ 2005GL024379.
- Wang, T., C. C. Y. Hung, and D. J. Randall. 2006. The comparative physiology of food deprivation: from feast to famine. Annual Review of Physiology 68:223–251.
- Watson, M. J. O., and A. A. Hoffmann. 1996. Acclimation, crossgeneration effects, and the response to selection for increased cold resistance in *Drosophila*. Evolution 50:1182–1192.
- Wegner, K. M., M. Kalbe, M. Milinski, and T. B. H. Reusch. 2008.

Mortality selection during the 2003 European heat wave in threespined sticklebacks: effects of parasites and MHC genotype. Evolutionary Biology 8:124.

- Welbergen, J. A., S. M. Klose, N. Markus, and P. Eby. 2008. Climate change and the effects of temperature extremes on Australian flying foxes. Proceedings of the Royal Society B: Biological Sciences 275: 419–425.
- Williams, A. P., C. J. Still, D. T. Fischer, and S. W. Leavitt. 2008. The influence of summertime fog and overcast clouds on the growth of a coastal California pine: a tree-ring study. Oecologia (Berlin) 156:601–611.
- Williams, S. E., L. P. Shoo, J. L. Isaac, A. A. Hoffmann, and G. Langham. 2009. Towards an integrated framework for assessing the vulnerability of species to climate change. PloS Biology 6:e325, doi:10.1371/journal.pbio.0060325.
- Zhou, L., A. Dai, Y. Dai, R. S. Vose, C.-Z. Zou, Y. Tian, and H. Chen. 2009a. Spatial dependence of diurnal temperature range trends on precipitation from 1950 to 2004. Climate Dynamics 32:429–440.
- Zhou, L., R. E. Dickinson, P. Dirmeyer, A. Dai, and S.-K. Min. 2009b. Spatiotemporal patterns of changes in maximum and minimum temperatures in multi-model simulations. Geophysical Research Letters 36:L02702, doi:10.1029/2008GL036141.
- Zimmermann, N. E., N. G. Yoccoz, T. C. Edwards Jr., E. S. Meier, W. Thuiller, A. Guisan, D. R. Schmatz, and P. B. Pearman. 2009. Climate extremes improve predictions of spatial patterns of tree species. Proceedings of the National Academy of Sciences of the USA 106:19723–19728.

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Despite the high diversity of ecological and life-history characteristics among squamate reptiles, climate variables such as rainfall have had strong influences on the thermal performance of this group. Some exemplars: top left, Cape dwarf chameleon *Bradypodion pumilum*; top right, Southern rock agama *Agama atra*; bottom left, rock monitor *Varanus albigularis*; bottom right, Ovambo tree skink *Trachylepis binotata*. Photographs by Susana Clusella-Trullas.