Title: Endotherms trade body temperature regulation for the stress response

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Abstract

Responding to perceived threats is energetically expensive and can require animals to curtail somatic repair, immunity, and even reproduction to balance energy ledgers. Among birds and mammals, energetic demands of thermoregulation are often immense, yet whether homeostatic body temperatures are also compromised to aid the stress response is unknown. Using data sourced from over 60 years of literature and 24 endotherm species, we show that exposure to non-thermal challenges (e.g. human interaction, social threats) caused body temperatures to decrease in the cold and increase in the warmth, but particularly when species-specific costs of thermoregulation were high and surplus energy low. Biophysical models revealed that allowing body temperature to change in this way liberated up to 24% (mean = 5%) of resting energy expenditure for use toward coping. While useful to avoid energetic overload, such responses nevertheless heighten risks of cold- or heat-induced damage, particularly when coincident with cold- or heat-waves.

Keywords: Trade-offs, thermoregulation, energetics, climate change

Introduction

About 2000 years ago, the Greek physician Galen observed that the body temperatures of his patients tended to rise after psychologically distressing experiences [1]. Given the acute nature of these rises, Galen termed their occurrence "ephemeral fevers" [2] and speculated, like those after him (e.g. Ibn Sina) that they emerged from an abundance of humor [3,4]. "Why" (i.e., functionally) they might emerge at all, however, was apparently left unquestioned.

Numerous studies have now shown that Galen's "ephemeral fever" is not just restricted to humans. Instead, many rodent, non-hominid primate, and even bird species appear to change their body temperatures after exposure to common disturbances, such as transportation or restraint [5-7]. Moreover, accumulating evidence suggests that the term "ephemeral fever" is likely a misnomer, with body temperatures of some individuals *declining* in response to some stressors rather than rising [8,9]. That these thermal responses to stressors exist across varied taxonomic groups and may differ under certain conditions strongly intimates a possible function, or, perhaps, a conserved and inherent nature (i.e., a constraint) behind their occurrence.

Recently, we proposed that changes in body temperature following stress exposure may be understood as consequences of an allocation trade-off between thermoregulation and the stress response ([10-11]; similar to ref. [12]). Under this hypothesis, fluctuations in body temperature that accompany the stress response represent the outcomes of reducing energy expenditure toward thermoregulation in favor of expenditure toward coping. Unlike previous hypotheses, this particular hypothesis provides an explanation for why body temperature might either rise *or* decline after a stress exposure, since relaxing thermoregulation should merely result in body temperature drifting toward ambient (particularly in the cold) or in the direction of accumulating metabolic heat (i.e. from activation of the stress response and reduced windows for heat loss; Fig. 1). Over the past few decades, a handful of studies have indeed supported a role of ambient temperature in

shaping the direction and magnitude of stress-induced thermal responses [13-15].

Moreover, others extending over several decades have already shown that

thermoregulatory investment can well be shifted to cope with other environmental

challenges, such as water deprivation (most notably in dromedaries, *Camelus dromedarius,*

ref [16]) and food limitation [17-18]. Together, these findings provide intriguing, albeit

sometimes conflicting, support for our trade-off hypothesis. Nevertheless, whether an

influence of ambient temperature holds across studies, species, and contexts (i.e.

stressors), remains unknown, but would have importantly implications for understanding

how, and at what cost, endotherms cope with non-thermal challenges.

Here, we used data describing body temperature responses to stressors across 24

terrestrial endothermic species to test for an allocation trade-off between

thermoregulation and the stress response (outlined in Fig. 1). Non-terrestrial endotherms

were excluded from our study owing to a lack of available data, but implications with

respect to these species are discussed. Under our trade-off hypothesis, we made three

predictions. First, although body temperatures should typically increase after a stressor

owing to accumulation of metabolic heat, these increases should be exacerbated in the

warmth and forgone, or even reversed, in the cold. Second, the degree to which ambient

temperature influences stress-induced thermal responses should depend on body mass,

with smaller species/individuals becoming more heterothermic across observed

temperature ranges, owing to higher surface area to volume ratios and greater

susceptibilities to the cold (i.e. higher lower critical temperatures and lower thermal

inertia). Third, how stress-induced thermal responses change across ambient temperature

should depend on the relative amount of energy available for allocation toward both

thermoregulation and the stress response (as per ref. [19]). More specifically, species with

relatively high metabolic rates (thus indicating low energy availability; see ref. [20]) should

be more likely to decrease their body temperature in the cold, and increase their body

temperature in the warmth following a stressor than those with relatively low metabolic

rates (indicating high energy availability).

Our findings indicate that body temperature regulations can be, and often is, compromised

by non-thermal challenges among bird and mammal species. These findings raise

important questions about risks posed by non-thermal stressors (e.g. human interaction)

experienced during increasingly-common extreme climate events. Further still, we argue

that they add to our growing understanding of how thermoregulatory strategies may be

altered to avoid homeostatic disruptions or states of energetic overload in challenging

environments (e.g. refs [16-18]).

Methods
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Literature search
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We searched primary literature for data describing the magnitude and context of stress-

induced changes in body temperature across endotherms using Web of Science™. To

simplify interpretations of our findings, we considered only literature pertaining to

terrestrial endotherms, excluding humans (owing to confounding effects of clothing on

costs of thermoregulation).

Query terms for our search were: "stress", "body temperature" and "NOT heat

stress". In total, 3022 studies were returned by our query. From those returned, we

scanned abstracts for evidence that: (1) body temperature responses to stress exposure

were indeed measured, and (2) the species in which body temperature responses were

measured was endothermic, terrestrial, and non-human; those with abstracts lacking such

evidence were removed from consideration. Next, because stress-induced changes in core

body temperature rather than surface body temperature are: (1) arguably most relevant to

whole animal energy expenditure and (2) subject to most investigation over the past

century, studies describing stress-induced responses at surface body temperatures were

also omitted. From all remaining studies (Supplemental Fig. 55), we scanned reference lists

for relevant literature that had not been returned by Web of Science. Again, abstracts of

these studies were reviewed for evidence that stress-induced changes in core body

temperature were indeed measured and were retained or excluded accordingly (n = 31

studies added; Supplemental Fig. 55). Together, all searches and data collection were

conducted between 28 April and 27 July, 2020, with publication dates in our data-set

ranging from March, 1958 [21] to January, 2020 ([22]; mean publication year [rounded to
nearest integer] ± s.e.m. = 2003 ± 0.879).

Data extraction and inclusion criteria

From each study identified for possible inclusion (n = 136; Fig. 1), we harvested the
following data:

(1) The average stress-induced change in body temperature across a sample
population (Δ°C; ± s.d. or s.e.m). In some studies, stress-induced thermal responses
were monitored and reported across multiple study groups (i.e. per sex, air
temperature treatment, discrete time period, etc.). In these cases, we chose to treat
data from each group as an independent observation nested within a study. If
multiple measurements of change were reported for a given study group, we
collected only the value demonstrating the greatest increase or greatest decrease
relative to baseline body temperature for that group.

(2) The number of individuals measured.

(3) The study species and family.

(4) The age class of the sample population (factorial; adult, juvenile, or mixed,
according to age of reproductive maturity). If age class was not reported, we
assumed that the study population was comprised solely of adults.

(5) Sex of individuals in sample population (factorial; female, male, or mixed). If sex
was not reported, we assumed it to be mixed.

(6) The season of study, categorized as spring, summer, fall, or winter. If the season
of study was not provided, we set it as "not available (NA)".

(7) The average mass of individuals in the study population (grams, ± s.d. or s.e.m;
s.e.m. ≈ 0 [0.0001] if missing). If a range of masses was provided (i.e. in n = 220
observations across n = 84 studies), we calculated the theoretical median and used
it in place of average mass, with the standard deviation equaling the mass range

divided by 4. Where average mass of individuals was not reported, we extracted

mean values for the species, sex, and age-class of interest (if provided; see

categories 4-5 above) from primary literature (n = 285 observations across n = 43

species groupings and n = 117 studies). Categories (4) and (5) describe our

approach when data on sex and age-class were missing.

(8) Air temperature during study (in °C, ± s.d. or s.e.m). For field studies lacking

measures of air temperature, we obtained average values from nearby weather-

stations (< 50 km from the study location) across the reported duration of study (i.e.

according to specified date and time ranges) where available (n = 13 observations

from n = 3 studies; indicated in data files). Studies reporting stress-induced thermal

responses in media other than air were discarded and those that did not report

variation around air temperature means were assigned an s.e.m. of ≈ 0 (0.0001; n =

57 observations across n = 28 studies).

(9) Latency between pre-stress exposure and post-stress exposure temperature

measurements (seconds). Latency values greater than 8 hours (28800 seconds; > 3

s.d. above the mean [3544 seconds]) were excluded from our analyses (n = 4

observations).

(10) Technique used to measure body temperature (intra-peritoneal telemetry or

thermocouple [rectal/cloacal/throat]).

(11) Whether pharmaceuticals other than saline were administered to study

animals before or during stress exposure (binomial; true/false).

For most studies, we extracted these data from the article text or from article figures using

the open-source software FIJI [23]. For all remaining studies, data were requested from

article authors directly. Once data collection was complete, all studies lacking data from

categories 1-4 and 8-10, and/or satisfying category 11 were discarded. Together, our final

sample included 165 observations across 68 studies, 24 species, and 30 species strains (n =

12 avian; n = 18 mammalian; average number of observations/species ± s.d. = 6.98 ± 19.00;

phylogeny provided in supplemental information). Final samples sizes and summaries of

sample exclusions following the preferred reporting items for systematic reviews and
meta-analyses (PRISMA) are outlined in Supplemental Fig. 55.

Types of stressors applied across our sample were varied and numerous, ranging
from transportation (n = 4 studies), social defeat (n = 10), restraint (mechanically [n = 31
studies] or by handling [n = 43 studies]), exposure to a novel environment (n = 32 studies),
and others.

Metabolic rate data

We were interested in testing whether a species' relative energetic expenditure at rest (and
thus, their remaining energy available for allocation) influences the magnitude of its stress-
induced change in body temperature. To calculate relative energetic expenditure for
species or species strains included in our data set (in publications that met our inclusion
criteria: see "Data extraction and inclusion criteria" above), we again used Web of Science™
to search for primary literature reporting either their basal or resting metabolic rates
("BMR" and "RMR" respectively). Measures of metabolic rate within each study were
categorized as "basal" according to whether they were: (1) collected at thermoeneutrality,
(2) collected post absorption, and (3) obtained during the circadian rest phase of the target
species (i.e. criteria 1-3 from McKechnie and Wolf [24]). Measures that satisfied only
criteria 1-2 were categorized as "resting" and all others were discarded. To enhance our
search for metabolic data, we replicated our queries using Google Scholar™.

Since both BMR and RMR likely vary across seasons, we sought to obtain metabolic
data relevant to the season at which stress-induced changes in body temperature were
observed (again, see "Data extraction and inclusion criteria" above). In most cases,
however, the season in which metabolic expenditure and/or the stress-induced thermal
response were measured were not reported (n = 66 studies of 68). Moreover, to our
knowledge, some species lacked any primary data on basal or resting metabolism at all.
Thus, where necessary, measures of BMR and/or RMR were obtained from available, but
potentially mismatched, seasonal studies, from the most closely related species studied (n
= 5 species; relatedness determined using a consensus phylogeny approach described
below), or from order-specific allometric relationships (n = 2 species; 1 chiropteran species

and 1 passerine species). Following collection, all metabolic data were converted to a

common metric (watts, "W") using conversion coefficients described in the appendix.

To account for effects of body mass on energy expenditure across species, we

obtained ordinary residuals from a Bayesian linear mixed effects model (LMM) predicting

natural log-transformed BMR or RMR as a function of natural log-transformed species mass

(mean in g; prior = $\mathcal{N}$[0.67, 0.1]). Type of measurement was included in this model as a

group-level intercept (i.e. "basal" or "resting"; group-level predictor; prior = Γ[1, 1]; model

coefficients: Intercept = -3.527 [-4.508, -1.873]; log mass = 0.668 [0.599, 0.734];

measurement type = 0.084 [-0.950, 0.869]) but phylogeny was not included to avoid

redundancy in our subsequent analyses. Residuals from our model were then categorized

as either positive or negative, with positive values for a given species indicating a relatively

high energy expenditure at rest, and negative values indicating a relatively low energy

expenditure at rest. Categorizing data in this way allowed us to broadly and conservatively

evaluate how relative energy expenditure, and thus, energy availability, influenced stress-

induced thermal responses while circumventing over-parameterization of our statistical

model (described below).

Effect size calculation

To ensure that our data were statistically comparable among studies and species, we

converted our estimates of absolute changes in body temperature to relative changes in

body temperature. This was achieved by calculating a response ratio as follows:

$$(1)\frac{T_b1_i}{T_b0_i}$$

where T_b1 represents an average minimal, or an average maximal body temperature of

study population i that was observed after a stress exposure, and T_b0 represents the body

temperature of study population i before a stress exposure treatment. For studies that also

recorded the body temperatures of control (or "unstressed") individuals throughout

experimentation, however, we used the following modified equations:

$$(2)\ \frac{T_b 1_{Si}}{T_b 1_{Ci}}$$

where T_b1_S again represents an average minimal, or an average maximal stress-induced
body temperature of the stress-exposed population *i*, and T_b1_C represents the average body
temperature of the control population at the equivalent time point.

To estimate uncertainty around our response ratio values, we calculated a pooled
standard error using one of the following equation:

$$(3)\ \sqrt{\frac{\sigma_{T_b'1}}{T_b 1} + \frac{\sigma_{T_b'0}}{T_b 0}}$$

or:

$$(4)\ \sqrt{\frac{\sigma_{T_b'1_S}}{T_b 1_S} + \frac{\sigma_{T_b'1_C}}{T_b 1_C}}$$

where equation (3) applies to studies without a control population, equation (4) applies to
studies with a control population. In these equations, σ_x represents the standard error of a
given subscripted variable *x*, and all other variables remain as previously described. In
studies where a standard error around the stress-induced change in body temperature was
reported, we assumed it to be a pooled estimate and used it in place of that derived from an
equation above.

Finally, to both normalize our response ratio estimates and centre them on zero, we
calculated their natural logarithms and used these transformed values as our final
estimates of effect size (i.e. the log-response ratio, or "lRR"). Effect sizes for studies with
and without control populations were visually compared to confirm that their distributions
were similar.

Phylogeny

To estimate phylogenetic relationships among species included in our data-set, we extracted two sets of 1000 trimmed phylogenetic trees – one for mammals, and one for birds – from VertLife [25] and BirdTree [26] respectively. These tree-sets were drawn from pseudo-posterior distributions of relaxed-clock models to infer phylogeny from genetic and fossil data (number of genes: mammals = 31[25]; birds = 15 [26]; avian fossil backbone from ref. [27]). From each tree-set, we then estimated class-specific consensus trees using "TreeAnnotator" in Beast 2 [28], with a 25% burn-in, posterior probability limit of 1.0, and with node heights representing median values. Once our consensus trees were established, we used them to build variance-covariance matrices approximating the relative relatedness between species (via the R package "ape" [29]; matrices used in subsequent analyses). For simplicity, relatedness between all pairs of birds and mammals was assumed to be negligible and set to zero. Phylogeny estimation and consensus trees are detailed in the appendix (section 5).

Statistical analyses

All of our statistical analyses were conducted in R version 4.03 [30] and figures were produced using the R package "ggplot2" [31]. For each model, we used Hamiltonian Monte Carlo (HMC) sampling in the R package "brms" [32] with 2500 burn-in iterations, 15000 sampling iterations, and a thinning factor of 10 across four HMC chains. Models were validated by visually diagnosing residual distributions and paired posterior coefficient estimates. Gelman-Rubin statistics ($\hat{R}$) for each of our models fell between 0.95 and 1.05, suggesting that chain mixing was consistently adequate, and effective sample size to sample size ratios consistently exceeded 0.75, confirming minimal autocorrelation between HMC chain draws. Versions for all R packages used are reported in the appendix.

To confirm that our results were not biased by small sample sizes among our collated studies, we re-ran of our model using an adjusted log response ratio designed to correct for sample-size related error ("RR$^{\Delta}$"; [33]). Results from this model were then compared with our original model at 80% and 95% highest density credible intervals (HDIs). Outcomes of both models were highly similar (see supplemental information; Table

S24), suggesting that sample variation in sample size among our collected studies was
unlikely to influence our reported findings.

Modeling

We were interested in testing whether the direction and magnitude of stress-induced
changes in body temperature depended on an individual's thermoregulatory demand
(described below) and relative energy available for investment in thermoregulation and
the stress response (here, residual RMR after controlling from body mass, categorized as
positive or negative). Clearly, thermoregulatory demand should vary according to air
temperature, with the degree of variation being influenced by a species' body size (here,
proxied by body mass). Further, the relative energy available to a species is predicted to be
inversely proportional to it's relative expenditure (i.e. at rest; see [20]), barring other
homeostatic compensations (refer to Discussion). Accordingly, we tested whether air
temperature, body mass, relative energy expenditure, and the interactions between both
air temperature and body mass, and air temperature and relative energy expenditure
predicted the direction and magnitude of stress-induced changes in core body
temperature. To achieve these ends, we constructed a Bayesian linear mixed effects model
(LMM) with our log response ratio as the response variable, and air temperature (°C), the
natural log of body mass (g), relative energy expenditure (factorial; positive or negative,
indicating high or low expenditure for a given body mass respectively), the linear
interaction between air temperature and body mass, and the linear interaction between air
temperature and relative energy expenditure as population-level predictors. Given that
both body mass and air temperature can vary considerably within studies, however, we
modeled each parameter as latent, with their true values being unknown but falling within
our observed values ± our observed error. Finally, both air temperature and the log of body
mass were mean-centered to better facilitate interpretation of our model intercept.

To account for recovery or ceiling/floor effects in our analysis, we included the
latency between baseline and stress-induced body temperature measurements as a
population-level predictor in our LMM. Here, we assumed that the temporal effects of
stress exposure on body temperature were non-linear, with body temperature rapidly

changing, then plateauing or recovering across time [34-36]. For this reason, the latency
between baseline and stress-exposed measurements was modeled as a second-order
polynomial. Similar to air temperature and the log of body mass, latency between
measurements was mean-centered to ease interpretation of our intercept. While the
duration of a stressor may also influence the amount of energy allocated toward the stress
response, estimates of this duration were correlated with our latency variable that was
already included in our model (Bayesian LMM: Intercept = -349.46 [-688.85, 3.23]; effect of
stressor duration [s; mean-centred] = 0.18 [0.09, 0.26]; study identity [group-level
intercept] = 1533.48 [1231.85, 1852.27]; priors: Intercept = $\mathcal{N}$[0, 300]; stressor duration =
$\mathcal{N}$(0, 10); study identity = exponential[0.01]; n = 149). Further, measuring stressor
duration is often unclear (e.g. for stressors such as social defeat). As such, stressor duration
was not included in our analysis. Finally, to correct for an influence of the method used to
measure body temperature on stress-induced thermal responses, and to correct for the
possibility of age effects, we included measurement technique (factorial; telemetry or
thermocouple) and age of individuals in the study population (integer; juveniles only = -1,
both juveniles and adults = 0, adults only = 1) in our model as population-level predictors.
Species, study identity, and phylogeny (see section 6) were included as group-level
predictors to control for non-independence between each factor.

For our LMM, we used weakly informative priors that assumed that log-responses
ratios of ⪅ -1 and ⪆ 0.5 (e.g. representing a 24°C decrease and 10°C increase in body
temperature respectively at baseline temperatures of 38°C) were unlikely. Specifically,
priors for air temperature, log body mass, and the first-order effects of latency to body
temperature measurement were set as $\mathcal{N}$(0, 0.1), while those for the interaction between
log body mass and air temperature, and the interaction between relative energy
expenditure and air temperature were set as $\mathcal{N}$(0, 0.05). Because we were less certain
about the range of possible age-related and second-order time-related effects, priors for
age class and the second-order effect of time were set as $\mathcal{N}$(0, 0.25). Next, because both
relative energy expenditure and measurement methodology were factorial, priors for these
parameters was set as skew-normal ($\mathcal{SN}$) with ξ= 0, ω= 0.1, and α = -2.5. We chose to use a
$\mathcal{SN}$ distribution to capture the increased likelihood of observing negative response ratios

than positive response ratios (an inherent property of natural logarithm transformation).
Importantly, however, a $\mathcal{SN}$ distribution was not assumed for priors of continuous
predictors where absolute slope estimates were expected to be small. Lastly, priors for all
group-level predictors, and σ were weak and flat (-∞ - ∞), or gamma-distributed (group-
level predictors: Γ[1.5, 1.0]; σ: Γ[1.0, 1.0]).

Four observations (out of 165; ~2.5% of our data) were removed from our analysis,
owing to their abnormally large effect sizes relative to the mean (> 3.5 times the sample
s.d.; absolute change in body temperature > 10°C). To account for the possibility that
posterior distributions for our model coefficients were non-normal, we calculated modes
for all coefficients in place of means, and highest posterior density credible intervals (80%
and 95%) around these modes [37]). All calculated means were marginalized unless stated
otherwise, and absolute changes in body temperature were calculated from log-response
ratios and assumed a baseline body temperature of 38°C.

Where reported, positive log-response ratios represent a stress-induced increase in
body temperature, while negative log-response ratios represent a stress-induced decrease.

Estimating change in energy expenditure attributed to the stress-induced thermal response

A critical assumption of our trade-off hypothesis is that, in response to a stressor,
allowing body temperature to decrease in the cold or increase in the warmth liberates a
meaningful amount of energy that can then be allocated toward coping with that stressor.
With this assumption in mind, we sought to estimate the relative extent to which stress-
induced changes in body temperature reduced expenditure toward thermoregulation
among our sampled vertebrates. To do so, we first estimated resting metabolic rate of
sampled animals at their known housing- and baseline body temperatures (henceforth
"baseline RMR") following established biophysical models [38], and by using the R package
nicheMapR [39]. These estimates represented energy expenditure required to maintain
thermal balance and assumed that: (1) baseline body temperature was held constant, (2)
loss of body heat to the ambient environment occurred by radiation, forced convection, and
free convection, but not conduction, and (3) the rate at which body heat was lost to the
ambient environment was inversely related to the resistance of both the core body itself

and the insulative layer of plumage or pelage [38]. For our estimates, we assumed that all

animals were ellipsoids with volumes (m^3) equal to their mass in g (body density = 1000

kg/m^3; [39]), and with semi-major axes being 1.2 and 1.3 times longer than semi-minor

axes for birds and mammals respectively. Further, we also assumed that: (1) pelage or

plumage depth of sampled animals was equal the logarithm of their body mass (g) in cm

(e.g. pelage depth = 1 and 2 cm for a 10 and 100 g mammal respectively), (2) thermal

conductivity of an animal's pelage or plumage was 0.0272 W/m-°C [38], and (3) wind

speed during sampling was negligible (0.1 m/s). Following these estimates, we then re-

estimated resting metabolic rates across the same ambient temperatures, but with body

temperatures representing those measured after stress exposure (henceforth, "stress-

induced RMR"), rather than at baseline. The difference between stress-induced and

baseline RMR values then represented the total energy saved or lost by adjusting body

temperature in the manner observed. These change values are expected to be conservative,

owing to the exclusion of temperature effects of metabolic processes themselves (i.e.

through q_{10} effects; [40]).

Results

Stressors trigger heterothermy and reduce costs of thermoregulation, particularly in small endotherms

Exposure to stressors coincided with, on average, an increase in body temperature of

0.64°C (indicated by our model intercept; $\beta \approx 1.67 \times 10^{-2}$ [80%: 4.41×10^{-3}, 2.73×10^{-2}; 95%: -

1.78×10^{-3}, 3.50×10^{-2}]). Both the magnitude and direction of these stress-induced responses

were clearly influenced by ambient temperature, as predicted by our trade-off hypothesis

($\beta \approx 7.70 \times 10^{-4}$ [80%: 3.10×10^{-4}, 1.23×10^{-3}; 95%: 4.00×10^{-5}, 1.44×10^{-3}]; Fig. 2). At low ambient

temperatures, where costs of warming are highest, body temperatures was more likely to

decrease after a stressor than at high ambient temperatures. By contrast, at high ambient

temperatures, where costs of warming are lowest and those of cooling highest, body

temperatures was more likely to increase after a stressor, and to a greater degree, than at

low ambient temperatures (Fig. 2).

An influence of ambient temperature on stress-induced responses was not

ubiquitous, and clearly depended on an individual's body mass (interaction between air

temperature and body mass: β = -4.00$\times$10^{-4} [80%: -6.40$\times$10^{-4}, -1.50$\times$10^{-4}; 95%: -7.80$\times$10^{-4}, -

1.00$\times$10^{-5}]; Fig. 3a). Given that body mass is capable of dictating both costs of

thermoregulation and susceptibilities to heat and cold among animals (i.e. the breadth of

their thermoneutral zones), this interactive effect between mass and ambient temperature

was expected under our trade-off hypothesis. Among individuals with low body masses (<

222 g, our sample mean), exposure to a stressor tended to induce a state of heterothermy,

with body temperatures decreasing after a stressor in cool temperatures and increasing

after a stressor in the warmth (Fig. 3a). For example, when ambient temperatures were

below our sample mean (20.1°C), body temperatures of these individuals fell by an average

of 0.54°C after a stress exposure, but when ambient temperatures were above our sample

mean, body temperatures of the same individuals rose by an average of 0.79°C (Fig. 3b).

Thermal responses of individuals with relatively high body masses (> 222 g), however,

were less nuanced. Specifically, body temperatures of these individuals tended to increase

after a stressor regardless of the current ambient temperature (Fig. 3; mean change in body

temperature = +2.02°C at below-average air temperatures, and +1.18°C at above-average

air temperatures). Regardless of body mass and ambient temperature, however,

biophysical models predicted that how body temperatures responded to stressors across

these animals reduced energetic costs of maintenance by on average of 4.8%, and up to

24% of resting metabolic rate (Fig. 3c).

Independent of air temperature, larger animals were marginally more likely to

increase their body temperature after a stress exposure, and by a greater degree, than

smaller animals (main effect of body mass: β = 4.25$\times$10^{-3} [80%: 1.26$\times$10^{-3}, 7.68$\times$10^{-3}; 95%: -

1.28$\times$10^{-3}, 9.17$\times$10^{-3}]; Fig. 2).

Limited surplus energy drives heterothermic responses to stressors

Body temperatures tended to increase after a stress exposure at moderate-to-high ambient

temperatures (indicated by a positive log response ratio at ambient temperatures >

approximately 10°C), regardless of whether a species expended relatively low, or relatively

high amounts of energy at rest (i.e. for their given mass; Fig. 4). The magnitude by which

these increases occurred also did not depend upon a species' relative energy expenditure

at rest (main effect of relative resting metabolic rate: $\beta \approx -3.98 \times 10^{-3}$ [80%: -1.34×10^{-2},

3.43×10^{-3}; 95%: -1.76×10^{-2}, 8.47×10^{-3}]; Fig. 2). In the cold (ambient temperatures <

approximately 10°C), however, species expending relatively high amounts of energy at rest

(and thus, having less energy available for allocation elsewhere) tended to decrease their

body temperatures, or allow them to fall, relative to those expending comparably little

energy at rest (interaction between air temperature and relative energy expenditure: $\beta =$

1.38×10^{-3} [80%: 2.60×10^{-4}, 2.23×10^{-3}; 95%: -2.70×10^{-4}, 2.79×10^{-3}]; Fig. 4a). This effect was

predicted by our trade-off hypothesis. Since baseline body temperatures did not differ

between species with relatively high, and relative low energy expenditure, this interactive

effect with ambient temperature could not be explained by body temperatures being near

maximum or minimum potential values alone (e.g. "ceiling" or "floor" effects; mean

baseline body temperature ± s.d.: low energy expenditure = 37.5°C ± 1.70°C; high energy

expenditure = 38.3°C ± 1.75°C).

Across all studies where ambient temperatures were below our sample average

(20.1°C), body temperatures of those with relatively high resting energy expenditure

decreased by an average of 0.37°C in response to a stressor (Fig. 4b). Among those with

relatively low resting energy expenditure, however, body temperatures increase by an

average of 0.74°C across the same ambient temperature range (Fig. 4b). In both cases,

biophysical models indicate that changing body temperature in these ways contributed to a

5% reduction (mean = 5.46%) in maintenance expenditure (Fig. 4c).

Time, age class, and species identity also influence the effect of stress exposures on thermal

physiology

Beyond air temperature, body mass, and relative energy expenditure, study methodology

had mixed effects on the outcomes of stress-induced thermal responses. Alone, latency

between baseline and stress-induced body temperature measurements clearly influenced

the observed magnitude of stress-induced thermal responses (first order: $\beta = -3.61 \times 10^{-2}$

[80%: -7.14×10^{-2}, -1.37×10^{-2}; 95%: -8.53×10^{-2}, 2.50×10^{-3}]; second order: $\beta = -5.04 \times 10^{-2}$ [80%:

-7.50×10^{-2}, -1.96×10^{-2}; 95%: -8.99×10^{-2}, -3.34×10^{-3}]; Figs. 2 & Supplemental Fig 56). Across

all studies assessed here, and when holding all other parameters constant, body

temperatures tended to increase immediately after stress exposure, and then decrease

after an average of approximately 35 minutes (2100 seconds; Supplemental Fig. 56).

Whether body temperature was measured using implanted telemeters or thermocouples,

however, did not appear to influence how body temperatures changed in response to

stressors ($\beta = 4.80 \times 10^{-4}$ [80%: -5.45×10^{-3}, 8.66×10^{-3}; 95%: -9.32×10^{-3}, 1.24×10^{-2}]; Fig. 2).

Although not predicted here, body temperatures were more likely to decrease after

a stress exposure among study populations comprised of solely adults than among those

comprised of solely juveniles ($\beta = -6.97 \times 10^{-3}$ [80%: -1.25×10^{-2}, -1.37×10^{-3}; 95%: -1.62×10^{-3},

1.39×10^{-3}]; Fig. 2). How body temperature responded to stress exposures was also weakly

genetically linked, with species identity and phylogeny explaining a small degree of

variation in our data (group-level intercept for species: $\mu = 1.49 \times 10^{-2}$ [80%: 7.55×10^{-3},

2.94×10^{-2}; 95%: 4.00×10^{-4}, 3.22×10^{-2}]; group-level intercept for phylogeny: $\mu = 1.50 \times 10^{-3}$

[80%: 0, 2.23×10^{-3}; 95%: 0, 3.25×10^{-3}]; Fig. 2).

Discussion

Costs of thermoregulation dictate body temperature under stress

Changes in body temperature that accompany the stress response are often understood as

byproducts of the physiological stress response and serve no function in and of themselves

[41-43]. Our results do not support this understanding. Instead, we provide evidence that

stress-induced changes in body temperature result from attempts to balance energetic

expenditure between body temperature regulation and the stress response. More

specifically, stress-induced thermal responses represent allocation trade-offs between

thermoregulation and stress physiology [10,11].

Supporting our trade-off hypothesis, our findings indicate that, among terrestrial

endotherms, how body temperature changes after a stress exposure may depends on how

much energy a species expends on body temperature regulation (as influenced by their

body mass, and thus, their surface area to volume ratio). In cool environments, for example,

species predicted to have relatively high expenditure toward warming (i.e. those with low

body mass, and thus, high surface area to volume ratios) tended to decrease their body

temperatures, or allow them to fall, following a stress exposure, while those with relatively

low expenditure toward warming, or even slight expenditure toward cooling (i.e. those

with large body masses, and thus, low surface area to volume ratios) did not (Fig. 3). Under

a trade-off hypothesis, we predicted this to be the case because expenditure toward the

stress response should only necessitate reductions in expenditure toward

thermoregulation when costs of warming or cooling are already high. By contrast, in the

absence of a trade-off, expenditure toward thermoregulation should be irrelevant to the

direction or magnitude of stress-induced thermal responses.

An interactive effect of ambient temperature and body mass on the outcome of

stress-induced thermal responses is not new to physiological theory. In birds, Nord and

Folkow [44] highlighted that hypothermic responses to stress exposure seem more likely

to occur at low ambient temperatures, but specifically among species with small body size

(e.g. Parids; [12,45]). Oka [46] alluded to a similar trend in small mammals (e.g. rats and

domestic rabbits; [13,14,47]). For Nord and Folkow [44], this interaction between body

mass and ambient temperature was interpreted as a byproduct of experimental

methodology, with commonly-used restraint stressors causing greater disruptions to the

retention of body heat in small birds relative to large birds (i.e. by compression of

plumage). Although this interpretation may be valid in some contexts, it alone does not

seem sufficient to explain our general findings. Numerous studies included in our analysis

(n = 28), employed stressors that did not involve direct contact or handling of

experimental animals at all (e.g. social isolation, exposure to a dominant individual,

introduction to a novel environment etc). Yet despite inclusion of these studies in our

analysis, we still detected a strong interactive effect of body mass and ambient temperature

on the outcome of stress-induced thermal responses (Fig. 3). Thus, interpreting these

responses as passive, unregulated consequences of experimental design is likely

premature.

Under a trade-off hypothesis, we speculated that body mass influenced stress-

induced thermal responses through its effects on heat loss and, subsequently, energetic

demands for heat production (discussed above). Nevertheless, it remains possible that how

body mass influences thermal responses to stressors occurs though other avenues, for

example, by influencing thermal inertia. Arguably the most direct way to test our

speculation would be to compare how body temperature responds to stressors at ambient

temperatures below, within, and above the zone of thermoneutrality within a given species,

where relative expenditure toward thermoregulation differs predictably. To the best of our

knowledge, very few studies have sought to conduct this type of comparison and outcomes

of those that have are equivocal [13,14,48].

Energy availability influences whether stressors induce heterothermy

Above and beyond body mass, our results indicate that relative energy expenditure

at rest influences how body temperature responds to stressors in different thermal

environment. Specifically, we found that species with relatively high rates of energy

expenditure at rest – whose energy available to allocate to both the stress response and

thermoregulation should be limited [20](but see below) – tended to lower their body

temperatures after stressors in the cold, but raise them in the warmth (Fig. 4). By contrast,

species with relatively low rates of energy expenditure at rest – whose energy available for

allocation toward the stress response and thermoregulation should abound – tended to

raise their body temperatures after stressors regardless of the thermal environment (Fig.

4). According to our trade-off hypothesis, we predicted this combined effect of energy

availability and the thermal environment on stress-induced thermal responses, since

adjustments of expenditure toward body temperature regulation should predominantly

occur when energy is limited [19].

For this study, we estimated that species with relatively high resting metabolism

(RMR) for their given mass would have less energy available for allocation than others (as

per [20]). While useful for broad inference, we recognize that this method for estimating

energy availability is coarse and ignores: (1) individual-level variation, and (2) the

possibility of high RMR species having other mechanisms in place to free, or acquire (but at

possible risk of predation) energy for allocation in the face of a challenge. Nevertheless,

studies directly or indirectly manipulating energy availability via food restriction also

support a trade-off hypothesis. In adult rats, for example, food restriction reversed thermal

responses to stressors, with fasted individuals decreasing their body temperatures in

response to restraint, and fed individuals increasing them in equivalent, sub-thermoneutral

conditions [49] (see ref [50] for range of thermoneutrality). Further, in birds, socially-

mediated food restriction (i.e. by social subordination) was shown to stimulate stress-

induced hypothermia in the cold and hyperthermia in the warmth (core body temperature

proxied by eye region temperature; [11]; similar findings in ref [51]). These findings

indicate that how body temperature changes after a stressor probably does depend on an

individual's relative energy balance, and may well represent active calibration of

expenditure toward warming or cooling. Further studies employing experimental food

would, however, be valuable to further clarify the role of energy balance in dictated stress-

induced thermal responses.

Not surprisingly, our results reiterate a key effect of time on the magnitude and

direction of stress-induced thermal responses. When controlling for all other parameters,

body temperature was more likely to increase shortly after stress exposure, then decrease

to baseline temperatures, or below, shortly thereafter (Fig. 2; Supplemental Fig. 56). These

trends strongly align with those reported for individual species [34-36] and probably

represent the timeline of physiologically responding to, and recovering from, a stressor.

Perhaps more interestingly, our results also indicated that age likely influences how body

temperature responds to a stressor. When averaged across species and thermal

environments, we found that younger individuals were more likely to increase their body

temperatures after a stressor, and to a greater extent, than older individuals (refer to

negative slope of age class in Fig. 2). Given that juveniles are expected to allocate high

amounts of energy toward growth [52,53], this result seems surprising. Although a similar

effect of age has been reported in a handful of other studies [49,54-56], why these

differences between juveniles and adults occur is not known. One possible explanation is

that juveniles are comparatively less averse to wear-and-tear associated with homeostatic

overload, once expenditure toward growth, thermoregulation, and the stress response are

combined (i.e., a "better safe than sorry" strategy; 56,57). Alternatively, the capacity to

change body temperature after a stressor may simply differ between age groups owing to

age-related changes to nervous system function (reviewed in ref. [46]). In any case,

understanding whether these age-related effects have meaningful implications on energy

balance is yet necessary to determine whether they support or refute our trade-off

hypothesis.

Putative mechanisms

Contextualizing stress-induced changes in body temperature as allocation trade-offs

(i.e. between thermoregulation and the stress response) may suggest that body

temperature set-point is actively changed when a non-thermal stressor is perceived.

Thermoregulatory control centers should therefore be involved in mediating these stress-

induced responses. So far, studies using pharmacological agents to manipulate neural

function in these control centers suggest that they probably do regulate body temperature

responses to stressors [57-60], at least indirectly. Additionally, their capacity to shape

stress-induced thermal responses is becoming increasingly clear [61,62]. In an effort to

synthesize these findings, Angilletta et al [63] recently proposed that inhibitory pathways

linking the limbic system with thermoregulatory nuclei of the hypothalamus may be

activated during the stress response, thus temporarily knocking down select

thermoregulatory processes. According to their model, activation of, and investment in, the

acute stress response is therefore actively prioritized over thermoregulation (representing

an "active" allocation trade-off). Efforts to map inhibitory pathways between limbic

structures and thermoregulatory control centers are limited but generally confirmatory

[64,65], thus providing some early mechanistic support for our trade-off hypothesis.

While changes to body temperature that accompany the stress response may well

represent "active" trade-offs, we still cannot rule out the possibility that they merely

represent passive allocation constraints. Namely, declines in body temperature in the cold

and increases in body temperature in the warmth might simply indicate a failing to allocate

energy toward responding to the thermal environment experienced during a stress

exposure. In this context, thermal responses to stressors may not constitute functional

responses to balance energetic expenditure between thermoregulation and the stress

response, but rather, may indicate an inability to regulate body temperature at all under

distress (see, for example, ref. [56]). Whether these stress-induced changes in body

temperature do indeed represent regulated trade-offs or passive constraints is an
important question that will directly influence whether these responses may respond to
selection or are simply fixed consequences of how physiological responses to stressors
shape body temperature.

Ectotherms and non-terrestrial endotherms

Descriptions of how body temperatures change after stressors among ectotherms
and non-terrestrial endotherms are sparse. We were therefore unable to formally evaluate
whether evidence for an allocation trade-off between thermoregulation and the stress
response extends to these species' groupings. However, indirect evidence for such a trade-
off in ectotherms is widely supportive. Some species of lizards, grasshoppers, and even
larval newts, for example, reportedly abandon basking and/or heat-seeking behavior in the
presence of predator cues (reviewed in ref. [63]). In the Yarrow's spiny lizard (*Sceloporus*
jarrovii), this change in behavior results in body temperature both declining toward
ambient and away from preference [66]. While not strictly evidencing an energy allocation
trade-off, these reports do highlight that responding to stressors can, and often does, take
precedence over body temperature regulation in ectotherms with clear consequences for
the latter. Among non-terrestrial endotherms, thermoregulation may also be relaxed, or
even forgone, in the presence of stressors. In southern bluefin tuna (*Thunnus maccoyii*), for
example, body temperatures can rise dramatically in response to suspected predators [67],
which, in severe cases, is thought to contribute to physiological damage ([68]; a
phenomenon known as "burnt meat"). We therefore speculate that our support for a trade-
off between coping was a stressor and regulating body temperature probably extends
beyond the species grouping analyzed here.

Conclusions

Our results provide strong evidence that changes in body temperature
accompanying the stress response represent allocation trade-offs between the stress-
response and thermoregulation. Across species, we highlight that by allowing body
temperature to drift toward ambient, or in the direction of accumulating metabolic heat,
these responses contribute to reductions in the energetic costs of thermoregulation,

particularly when such are already high, or when energy available for allocation to both the

stress response and body temperature regulation are low. These findings, therefore,

highlight a new model for understanding both how and "why" stress-induced changes in

body temperature might occur (outlined in Fig. 1). In doing so, they provide novel, cross-

species evidence that the acute stress response may directly influence body temperature

regulation in endotherms. Given the clear implication of these findings for a warming

world, further experimental studies testing our trade-off hypothesis within species should

be a priority in coming years.

 Figures

Figure 1. Trade-off model explaining the outcome of stress-induced changes in body temperature. A | "Available energy" indicates surplus energy available for allocation toward biological processes after accounting for maintenance metabolism (i.e. basal metabolism). Thickness of lines indicates the magnitude of influence that one variable exerts on another, in the direction of the arrow. **B |** Solid black lines indicate resting metabolism across ambient temperature for a hypothetical, terrestrial endotherm. Grey bands indicate ranges of ambient temperature where expenditure toward thermoregulation is expected. Solid grey lines indicate body temperature across ambient temperature for the same hypothetical endotherm during control (or unstressed) conditions and the dashed grey lines indicate body temperature across ambient temperature during stress-exposed conditions. Ribbons around grey lines are added to indicate that true body temperature responses to control and stress-exposed conditions will vary.

Figure 2. Factors influencing stress-induced changes in core body temperature among terrestrial endotherms. Dots represent the estimated effect of a given parameter on the magnitude and direction of stress-induced changes in body temperature (measured by a log response ratio, "lRR"). "Time" indicates latency to a maximum or minimum body temperature measurement after stressor onset; "relative resting metabolic rate" indicates the residuals of log-transformed resting metabolic rate (W) regressed against log-transformed body mass (g), then categorized as positive or negative. Negative slopes represent a tendency to reduce core body temperature, while positive slopes represent a tendency to increase core body temperature. Black grey represent 80% highest posterior density intervals credible intervals (HPDIs) and black whiskers represent 95% HPDIs.

Figure 3. Effect of air temperature and body mass on the magnitude and direction of stress-induced changed in core temperature. A. Negative log response ratios represent stress-induced declines in body temperature, while positive log response ratios represent stress-induced increases in body temperature. Trend lines represent predicted relationships for an animal of a given body mass (g), while holding all other model covariates (shown in Fig. 2) at their averages. Dark ribbons represent 80% highest posterior density intervals (HPDIs) and pale ribbons represent 95% HPDIs. **B.** Large dots indicate average changes in body temperature for species above and below mean body mass for our sample (~222g; beige and green respectively), and at ambient temperatures above and below the average across studies (20.1°C). Mean changes in body temperature are calculated from log-response ratios and assume a baseline body temperature of 38°C.

Small dots represent raw data values and error bars indicate 80% and 95% highest posterior density intervals (HPDIs; black and grey respectively). **C.** Large dots display mean changes in resting energy expenditure attributed to stress-induced thermal responses, as predicted by biophysical models. Again, small dots represent raw data values and error bars indicate 80% and 95% highest posterior density intervals (HPDIs; black and grey respectively).

Figure 4. Effect of relative energy expenditure at rest on the magnitude and direction of stress-induced changed in core temperature. Relative energy expenditure at rest represents the residuals extracted from a linear mixed effects model regressing log-transformed resting metabolic rate (W) against log-transformed body mass (g), then categorized as "low" (i.e. negative residuals) or "high" (i.e. positive residuals). **A.** Negative log response ratios represent stress-induced declines in body temperature, while positive log response ratios represent stress-induced increases in body temperature. Dark ribbons represent 80% highest posterior density intervals (HPDIs) and pale ribbons represent 95% HPDIs. Trend lines and credible intervals are marginalized across all other model predictors. **B.** Large dots represent average changes in body temperature for species with high or low relative energy expenditure (red and navy respectively), and at ambient

temperatures above and below the average across studies (20.1°C). Mean changes in body temperature were calculated from log-response ratios and assume a baseline body temperature of 38°C. Small dots represent raw data values and whiskers indicate 80% and 95% highest posterior density intervals (HPDIs) around means (black and grey respectively). **C.** Large dots display mean changes in resting energy expenditure attributed to stress-induced thermal responses, as predicted by biophysical models. Again, small dots represent raw data values and error bars indicate 80% and 95% highest posterior density intervals (HPDIs; black and grey respectively).

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Effect on Body Temperature
Responses to Stressors

Estimated Effect (Slope of lRR)