

## The decoupled nature of basal metabolic rate and body temperature in endotherm evolution

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2	evolution
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22 The origin of endothermy in birds and mammals are iconic events in vertebrate 23 evolution. Endotherms can maintain their body temperature  $(T_b)$  over a wide range 24 of ambient temperatures  $(T_a)$  using primarily the heat generated continuously by 25 their high basal metabolic rates  $(BMR)^1$ . There is also an important positive feedback loop in that *BMR* itself is influenced by  $T_b^{1-3}$ . Owing to this, many 26 27 ecologists and evolutionary physiologists posit that the evolution of BMR and  $T_{\rm b}$ 28 must have been coupled during the radiation of endotherms<sup>3-5</sup>, changing with 29 similar trends<sup>6-8</sup>. However, colder historical environments might have imposed 30 strong selective pressures on BMR to compensate for increased rates of heat loss and 31 to keep  $T_b$  constant<sup>9-12</sup>. Thus, adaptation to cold  $T_a$  via *BMR* increases could have 32 decoupled *BMR* from T<sub>b</sub> and caused different evolutionary routes to the modern 33 diversity in these traits. Here we show that *BMR* and  $T_{\rm b}$  were decoupled in ~ 90% of 34 mammalian and in ~ 36 % of avian phylogenetic branches. Mammalian BMR 35 evolved with rapid bursts but without any long-term directional trend, whereas  $T_{\rm b}$ 36 evolved mostly at a constant rate and towards colder bodies from a warmer-bodied 37 ancestor. Avian BMR evolved predominantly at a constant rate – again with no 38 trend, whereas  $T_{\rm b}$  evolved with much greater rate heterogeneity than *BMR* and 39 there has been adaptive evolution towards colder bodies. Furthermore, rapid shifts 40 leading to both BMR increases and decreases were linked to abrupt changes 41 towards colder  $T_a$  but only in mammals. Our results suggest that natural selection 42 effectively exploited the diversity of mammalian BMR under diverse, often adverse 43 historical thermal environments.

44

45 Phylogenetic statistical methods<sup>13, 14</sup> now provide us with the opportunity to formally test

46 whether *BMR* has been linked to  $T_b$  or  $T_a$  throughout the evolution of birds and mammals.

47 By accommodating for and identifying heterogeneity in the rate of phenotypic evolution

48 these methods can detect and reconstruct accurate historical evolutionary processes<sup>15</sup>.

49 Evaluation of the evolutionary coupling between *BMR* and  $T_b$  has direct consequences for

50 several longstanding ecological and evolutionary theories (including the Metabolic

51 Theory of Ecology) which assume  $coupling^{2-8}$ .

52

53 We first quantified and compared rates for *BMR* and  $T_b$  evolution along each branch of

54 the time-calibrated phylogenetic trees of birds and mammals (henceforth *branch-wise* 

55 rates, r; see Methods). r measures how fast a trait evolved along an individual

56 phylogenetic branch (r is a rate scalar by which the background rate is multiplied to 57 increase or decrease the pace of evolution). If *BMR* and  $T_b$  were coupled during the 58 evolution of endotherms, the amount of change along phylogenetic branches in both traits 59 should be positively associated – where r is high in BMR we expect it to be high in  $T_{\rm b}$ 60 (Fig. 1 b). We tested this prediction against alternative evolutionary scenarios. Firstly, we 61 cannot make any inferences about coupling or decoupling where there is no rate 62 heterogeneity for both *BMR* and  $T_b$  (r = 1 for all branches in the tree for both traits; Fig. 63 1a). Secondly, we infer decoupled evolution if both traits show rate heterogeneity, but the 64 magnitudes of rs are negatively correlated (*i.e.* branches evolving at a high rate for BMR 65 are evolving at a low rate for  $T_{\rm b}$ , and vice-versa, Fig. 1c). We suggest this scenario implies decoupled evolution because a negative correlation most likely implies that one 66 67 trait tends to be conserved whilst the other evolved rapidly. Thirdly, we infer decoupled 68 evolution if only one trait shows rate heterogeneity while the other evolved at a constant 69 rate (Fig 1d and e) or if both traits show heterogeneity but the *branch-wise rates* are not 70 associated (Fig 1f).

71

As *BMR*, body mass (*Mass*),  $T_b$ , and  $T_a$  are at least to some extent correlated in extant birds and mammals, and such correlations may vary between orders<sup>16</sup>, we estimated the *branch-wise rates* for *BMR* and  $T_b$  while accounting for their covariates across extant species using the *phylogenetic variable-rates regression* model<sup>17</sup> (henceforth *variablerates*; Methods). This approach allows for simultaneous estimation of both an overall relationship between, for instance, *BMR* as a function of *Mass* and  $T_b$  across extant species, and any shifts in rates (*r*) that apply to the phylogenetically structured residual

79	variance in the relationship. In both birds and mammals, the variable-rates model
80	significantly fits the data better than the constant-rate regressions, which assume a single
81	rate across all branches (Methods; Table S1 to S8). The best fitting variable-rates model
82	for mammalian <i>BMR</i> includes both <i>Mass</i> and $T_b$ with a single slope for each trait
83	estimated across all orders (Table S1 and S2). For mammalian $T_{b}$ , the best fitted model
84	includes Mass and BMR as covariates, also with a single slope across all orders (Table S3
85	and S7). In birds, the best model for BMR includes only Mass with a single slope for all
86	orders (Table S4). Finally, the best fitted model for avian $T_b$ includes Mass only in
87	Columbiformes (Table S6).
88	
89	The branch-wise rates estimated in the best fitting models shows that mammalian BMR
90	evolved at a constant rate ( $r = 1$ ) in just 11.2% of branches and at faster rates ( $r > 1$ ) in
91	88.8% of branches (Fig. 2a). Mammalian $T_b$ evolved at a constant rate in 70.3% of
92	branches and faster rates in 29.7% of branches (Fig. 2b). In birds, BMR evolved at a
93	constant rate in 90.5% of branches and at faster rates in 9.5% of branches (Fig. 2d).
94	Avian $T_{\rm b}$ evolved at a constant rate in 69 % of branches and at faster rates in 31% (Fig.
95	2e). When the <i>branch-wise rates</i> for <i>BMR</i> and $T_b$ were compared, we found that in
96	mammals, both traits evolved at a constant rate in 10.6% of branches (Fig. 3a consistent
97	with Fig. 1a). In 60.2% of branches only one trait evolved at faster rates while the other
98	trait diverged at a constant rate. This indicates that <i>BMR</i> and $T_b$ evolved in a decoupled
99	fashion along these branches (Fig. 3a consistent with Fig. 1d, e). We found that 29.2% of
100	branches had an increased rate in both $BMR$ and $T_b$ . However, the magnitudes of the
101	branch-wise rates were not significantly correlated ( $p_{MCMC}$ [% of posterior distribution

102	crossing zero] = 9%; Table S9; Fig. 3a consistent with Fig 1f). This also suggests
103	decoupled evolution in those branches – likely because of distinct selection pressures
104	acting on <i>BMR</i> and $T_{\rm b}$ . On the other hand, both traits evolved at a constant rate in 63.8%
105	of branches for birds (Fig. 3c consistent with Fig. 1a). In 32% of branches only one trait
106	evolved at fast rates while the other trait diverged at a constant rate (Fig. 3c consistent
107	with Fig. 1d, e). In the remaining 4.2% of branches, both traits evolved at faster rates, but
108	the <i>r</i> magnitudes were not statistically correlated ( $p_{MCMC} = 16.9\%$ , Table S10, Fig. 3c
109	consistent with Fig. 1f).
110	
111	As rapid bursts in <i>BMR</i> evolution were not coupled with those in $T_b$ evolution, we
112	evaluated the alternative hypothesis postulating that $BMR$ evolved in response to $T_a$ . This
113	hypothesis suggests that colder environments increase the rate of heat lost from
114	organisms which is subsequently compensated by $BMR$ increases <sup>9-12</sup> . These $BMR$
115	increases could have occurred over long periods of time because of global cooling <sup>18</sup> -
116	generating a long-term directional trend in BMR during the radiation of mammals and
117	birds. This expectation is in line with the Plesiomorphic-Apomorphic Endothermy
118	Model <sup>6-8</sup> (PAE Model). By assuming that <i>BMR</i> and $T_b$ are coupled in endotherms and that
119	they both can be used as a proxy of the degree of endothermy, the PAE model predicts a
120	general tendency towards higher endothermic levels through time (from basoendothermic
121	ancestors, Methods) associated with the Cenozoic global cooling. Global cooling is not
122	the only source of variation in $T_a$ . Long-term directional increases in <i>BMR</i> might have
123	also been driven by historical dispersals of endotherms towards higher latitudes <sup>19</sup> . In
124	either case, if a long-term decrease in $T_a$ drove adaptation via <i>BMR</i> elevation, and $T_b$

followed the same trajectory (as assumed by the PAE model) we expect to find a positive correlation between the *branch-wise rates* of *BMR* and the *branch-wise rates* of  $T_a$ . With this in mind, we also expect a positive trend towards higher *BMR* and  $T_b$  values from basoendothermic ancestors and a negative trend towards lower  $T_a$  from warmer ancestral environments. We used the *variable-rates* model to estimate the *branch-wise rates* for  $T_a$ whilst accounting for latitude since, generally,  $T_a$  decreases from the equator to the poles (Methods; Table S11).

132

133 The variable-rates model significantly improved the fit to the  $T_a$  data over the constant-134 rate regression model in both mammals and birds (Table S11). In 21.2% of mammalian 135 branches  $T_{\rm a}$  evolved at a constant rate, and with rate heterogeneity in the remaining 136 78.8% – including 72.2% of branches with faster rates and 6.6\% with slower rates (r < 1, 137 Fig. 2c). This indicates that most ancestral mammalian lineages (72.2%) faced abrupt 138 historical changes in their  $T_{\rm a}$ , while far fewer lineages (6.6%, mostly bats) survived and 139 continued existing in similar thermal environments. In birds, 77.6% of branches show 140 faster rates of  $T_a$  change, 22.1% show changes at a constant rate, and in only a single 141 branch the  $T_a$  changed at a slower rate (Fig. 2f). 142

143 When *branch-wise rates* of mammalian *BMR* and  $T_a$  evolution were compared, we found 144 that they were coupled in 74.9% of branches ( $p_{MCMC} = 0\%$ ; Table S12; Fig. 3b, consistent 145 with Fig. 1b). To evaluate further if  $T_a$  decreases were linked to *BMR* increases in the 146 74.9% of mammals where both traits were coupled (i.e. to ascertain the direction of 147 change), we evaluated the expected positive trend in *BMR* as a response to the long-term

148 decrease in  $T_{\rm a}$ . We conducted Bayesian phylogenetic regressions between extant values 149 of these two variables (in turn) and the *path-wise rates* (sum of *branch-wise rates* along 150 branches in the path from the root of the tree to each terminal species, Methods)<sup>15</sup>. We 151 found a negative effect of *path-wise rates* on T<sub>a</sub> across all mammals (Fig. 4b; Table S14), 152 which supports a long-term directional trend towards habitats with lower  $T_a$  over time. 153 However, we did not find evidence for any trend in mammalian BMR evolution – BMR154 increases and decreases were equally likely in our sample (Table S14). Our results 155 suggest that in colder environments, where resources were available to fuel metabolic elevation, selection favoured higher mammalian  $BMR^{20}$ . Another possibility might be that 156 157 *BMR* increase was a correlated response to direct selection on other physiological traits, 158 like maximal metabolic capacities for thermogenesis, whose benefits outweigh the energetic cost of BMR elevation<sup>20</sup>. Otherwise, selection may have always favoured BMR 159 decreases under an ever colder environment<sup>20</sup>. 160

161

162 In contrast to mammals, most avian branches that experienced rapid shifts in  $T_a$  did not 163 show evidence of coupled changes in BMR - 68.4% of branches had fast rates of  $T_a$ 164 evolution but a constant rate of *BMR* evolution (Fig. 3d consistent with Fig. 1d, e). 165 Moreover, the small fraction of branches where BMR evolved at fast rates (9.5%) were 166 not linked to rapid shifts in T<sub>a</sub> (Fig. 3d consistent with Fig. 1f; Table S13). Avian BMR 167 did not show a positive evolutionary trend despite the fact they also experienced colder 168 environments over time (Fig. 4d; Table S15). Birds might not have responded to colder 169 temperatures by changes in their *BMR* because their lower thermal conductance might

have helped them retain internal heat<sup>9</sup>. Alternatively, other physiological strategies, such
as torpor, may have been selected for under colder environments<sup>21</sup>.

173	Finally, we found a negative effect of <i>path-wise rates</i> on $T_b$ in both mammals (Fig. 4a;
174	Table S14) and birds (Fig. 4c; Table S15). This suggest that – on average – endotherms
175	evolved towards colder bodies from warmer-bodied ancestors. These directional models
176	predict a mean $T_{\rm b}$ of 35.3 °C and 40.4 °C in the most recent common ancestor (MRCA)
177	of mammals and birds respectively (Fig. 4a, c), suggesting that early birds and mammals
178	were mesoendotherm rather than basoendotherms (Methods). This result does not support
179	that ancestral mammals could not attain $T_{\rm b} > 30$ °C owing to the elevated metabolic rates
180	necessary to compensate heat loss in cold environments <sup>22</sup> . However, if the $T_b$ - $T_a$
181	differential ( $\Delta T$ ) determines how hot early mammals were, we expect that the mammalian
182	MRCA with a $T_b$ of 35.3 °C could survive in an environment warm enough to have a low
183	$\Delta T$ . Our model describing the negative trend in $T_a$ predicts that the MRCA of mammals
184	lived in an environment with 23 °C on average (Fig. 4b), resulting in a $\Delta T$ of 15.3 °C.
185	This ancestral $\Delta T$ is very conservative compared with the $\Delta T$ s observed in extant
186	mammals. For example, there are small mammals that achieve $T_b$ higher than 39 °C (e.g.
187	<i>Microdipodops pallidus</i> <sup>16</sup> ) that can survive in environments of 11 °C <sup>19</sup> ( $\Delta T = 28$ °C).
188	Also, some larger mammals have stable $T_b$ even in extreme environmental conditions –
189	the Artic hare ( <i>Lepus arcticus</i> ) can maintain its $T_b$ of 38 °C <sup>16</sup> in temperatures as low as -
190	12 °C <sup>19</sup> ( $\Delta T = 50$ °C).
191	

192	Taken together, our results reveal that $BMR$ was not coupled with $T_b$ across the evolution
193	of endothermic species. As environments became colder, mammals survived by changing
194	their BMR, while birds likely survived owing to their high thermal insulation. Evaluating
195	the isolated and/or combined effect of environmental variables on physiological attributes
196	has implications for evidence-based projections for the future <sup>23</sup> . In this sense, the
197	previously unappreciated complexity, interplay and decoupled nature in the evolutionary
198	history of <i>BMR</i> , $T_b$ and $T_a$ might point to undetected resilience of endotherms in the face
199	of modern global challenges.
200	
201	Figure legends.
202	
203	Figure 1. Possible evolutionary scenarios between $BMR$ and $T_b$ given their branch-
203 204	Figure 1. Possible evolutionary scenarios between <i>BMR</i> and $T_b$ given their <i>branch-wise rates</i> ( <i>r</i> ) in a bivariate space. Grey colours represent the constant background rate
204	wise rates $(r)$ in a bivariate space. Grey colours represent the constant background rate
204 205	<i>wise rates</i> ( <i>r</i> ) in a bivariate space. Grey colours represent the constant background rate $(r = 1)$ . Red colours represent rates faster than the background rate $(r > 1)$ and blue
204 205 206	<i>wise rates</i> ( <i>r</i> ) in a bivariate space. Grey colours represent the constant background rate $(r = 1)$ . Red colours represent rates faster than the background rate $(r > 1)$ and blue colours represent rates slower than the background rate $(r < 1)$ , which might be related to
204 205 206 207	<i>wise rates</i> ( <i>r</i> ) in a bivariate space. Grey colours represent the constant background rate $(r = 1)$ . Red colours represent rates faster than the background rate $(r > 1)$ and blue colours represent rates slower than the background rate $(r < 1)$ , which might be related to past events of positive <sup>17</sup> and stabilizing selection <sup>29</sup> respectively. Point fill colours
204 205 206 207 208 209	<i>wise rates</i> ( <i>r</i> ) in a bivariate space. Grey colours represent the constant background rate $(r = 1)$ . Red colours represent rates faster than the background rate $(r > 1)$ and blue colours represent rates slower than the background rate $(r < 1)$ , which might be related to past events of positive <sup>17</sup> and stabilizing selection <sup>29</sup> respectively. Point fill colours represent <i>BMR</i> rates and point outline colours represent $T_b$ rates.
204 205 206 207 208 209 210	wise rates (r) in a bivariate space. Grey colours represent the constant background rate $(r = 1)$ . Red colours represent rates faster than the background rate $(r > 1)$ and blue colours represent rates slower than the background rate $(r < 1)$ , which might be related to past events of positive <sup>17</sup> and stabilizing selection <sup>29</sup> respectively. Point fill colours represent <i>BMR</i> rates and point outline colours represent $T_b$ rates.
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214 Kee Yap (modified), Mathew Callaghan, Matt Martyniuk, nicubunu, Pearson Scott

- 215 Foresman, Prin Pattawaro et al., Rebecca Groom, Sarah Werning, T. Michael Keesey
- 216 (after Joseph Wolf), Yan Wong, Steven Traver. Silhouette licence links:
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- 218 https://creativecommons.org/licenses/by-sa/3.0/.
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220 Figure 3. *Branch-wise rates (r)* of *BMR*, *T*<sub>b</sub>, and *T*<sub>a</sub> in bivariate space for mammals

221 (a, b) and birds (c, d). Bayesian GLS analyses indicates that only fast *branch-wise rates* 

for *BMR* and slow-fast *branch-wise rates* of *T*<sub>a</sub> were statistically correlated in mammals

- 223  $(p_{MCMC} = 0; n = 602 \text{ branches; black line}).$
- 224

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Figure 4. Mammals (a, b) and birds (c, d) evolved towards both colder T_b and T_a
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226 over their evolutionary history. *Path-wise rates* had a significant negative effect in

mammalian and avian  $T_b$  ( $p_{MCMC} = 4\%$  and 3%; n = 502 and 367 species) and in

mammalian and avian  $T_a$  ( $p_{MCMC} = 0$  and 0; n = 2922 and 6142 species), both supporting

a negative macroevolutionary trend<sup>15</sup>. Transparent and dark lines indicate the posterior

- 230 distribution of slopes and the mean slope respectively, estimated from the Bayesian
- 231 PGLS (Methods).
- 232

## 233 Methods.

**Data**. We used a time-calibrated phylogenetic tree of extant mammals  $(n = 3321)^{24}$ , and the body mass (*M*), basal metabolic rate (*BMR*), and body temperature (*T*<sub>b</sub>) taken from

236 Clarke et al.<sup>16</sup> (n = 632). After identifying species in the tree that have trait information,

we obtained a final mammalian dataset of 502 species, which includes representativesfrom 15 orders (SI).

239

For birds, we used the consensus time-calibrated tree from Rolland et al.<sup>19</sup>. This tree was 240 inferred from the samples of trees provided by Jetz et al<sup>25</sup>. Data for *BMR*,  $T_{\rm b}$ , and *Mass* 241 were obtained from Fristoe et al<sup>9</sup>. After matching this database with the phylogenetic 242 243 tree, we obtained a final sample of 164 species which includes representatives from 21 244 orders (SI). The dataset used to evaluate evolutionary trends in  $T_b$  (see below) is from Clarke & Rothery<sup>26</sup>, which contains 367 species with phylogenetic information. 245 246 247 Data for ambient temperature  $(T_a)$  and latitude for extant mammals and birds was extracted from Rolland et al.<sup>19</sup> These datasets include 2922 species of mammals and 6142 248 species of birds which have phylogenetic information. The  $T_a$  for extant endothermic 249 250 species is the temperature of environments in which birds and mammals inhabit today -251 measured as the mean ambient temperature for the mid-point latitude of each species distribution (Rolland et al.<sup>19</sup>). The  $T_a$  at which a species exists today may not be a 252 253 heritable trait *per se*. However, the evolution of  $T_a$  can still be inferred using phylogenetic 254 methods since habitat selection reflects species adaptations (traits) to some characteristics 255 of the environment. This interrelationship should leave phylogenetic signal in the  $T_a$  at 256 which endothermic species live. Accordingly, we found significant phylogenetic signal in the  $T_a$  of both mammals ( $\lambda_{PosteriorMean} = 0.77$ ; Bayes Factor = 665) and birds ( $\lambda_{PosteriorMean} =$ 257 0.8; Bayes Factor = 1404). Furthermore, the phylogenetic signal for  $T_a$  is very high ( $\lambda$ =1) 258 in birds and mammals, when estimated using the median-*r* scaled tree. 259

Finally, to evaluate the endothermic levels for the MRCA of mammals and birds proposed by Lovegrove<sup>7, 8</sup>, we followed his categorization of endothermic species as basoendotherms ( $T_b^{\text{Birds}} < 40.4 \text{ °C}$ ;  $T_b^{\text{Mammals}} < 35.0 \text{ °C}$ ), mesoendotherma ( $40.4 \text{ °C} \le T_b^{\text{Birds}} \le 42.5 \text{ °C}$ ;  $35 \text{ °C} \le T_b^{\text{Mammals}} \le 37.9 \text{ °C}$ ), and supraendotherms ( $T_b^{\text{Birds}} > 42.5 \text{ °C}$ ;  $T_b^{\text{Mammals}} > 37.9 \text{ °C}$ ).

266

267 **Inferring the** *branch-wise rates* of evolution. We identified heterogeneity in the rate of 268 evolution along phylogenetic branches (*branch-wise rates*) by dividing the rate into two parameters: a background rate parameter ( $\sigma_b^2$ ) which assumes changes in the trait of 269 270 interest (e.g. *BMR*) are drawn from an underlying Brownian process, and a second 271 parameter, r, that identifies a branch-specific rate shift. A full set of *branch-wise rates* are 272 estimated by adjusting the lengths of each branch in a time-calibrated tree (stretching or 273 compressing a branch is equivalent to increasing or decreasing the phenotypic rate of 274 change relative to the underlying Brownian rate of evolution). Branch-wise rates are 275 defined by a set of branch-specific scalars  $r (0 < r < \infty)$  which transform each branch in order to optimize the phenotypic rate of change to a Brownian process ( $\sigma_{\rm b}^2 r$ ). If 276 277 phenotypic change occurred at accelerated (faster) rates along a specific branch of the 278 tree, then r > 1 and the branch is stretched. Decelerated (slower) rates of evolution are 279 detected by r < 1 and the branch is compressed. If the trait evolves at a constant rate 280 along a branch, then the branch will not be modified (*i.e.* r = 1). 281

282 We estimated the r values of BMR,  $T_{\rm b}$ , and  $T_{\rm a}$  evolution using the phylogenetic variable*rates regression* model in a Bayesian framework<sup>17</sup>. This model is designed to 283 284 automatically detect shifts in the rate of trait evolution across phylogenetic branches 285 while accounting for a relationship with another trait or traits across extant species 286 values. This approach allows for simultaneous estimation of both an overall relationship 287 between, for instance, BMR as a function of Mass and  $T_{\rm b}$  across extant species, and any 288 shifts in rates (r) that apply to the phylogenetically structured residual variance in the 289 relationship. As residual variance is explained by shifts in rate across phylogenetic 290 branches we can, for example, determine how much *BMR* has changed in the past (r)291 after accounting for their covariation with *Mass* and  $T_{\rm b}$  in the present (the relationship 292 between the values across extant species). Thus, if the amount of *BMR* change along individual phylogenetic branches were coupled with the amount of change of  $T_b$ , then we 293 294 should find the r values of BMR to be positively associated with the r values of  $T_{\rm b}$ . The 295 branch-wise rates for  $T_{\rm b}$  evolution can be estimated while accounting for its covariation 296 with other traits or factor across extant species. Previous studies on the association 297 between *BMR* and  $T_{\rm b}$  using extant species values alone have not evaluated the association 298 in evolutionary terms even when they use phylogenetic method.

299

300 We evaluated 24 *phylogenetic variable-rates regression models* and 24 *phylogenetic* 

301 *constant-rate regression models* (Table S1 to S8). Regression model selection was

302 conducted using Bayes Factors (BF) via marginal likelihoods estimated by stepping stone

303 sampling. BF is calculated as the double of the difference between the log marginal-

304 likelihood of the complex model and the simple model. By convention, BF > 2 indicates

305	positive evidence for the complex model, $BF$ 5-10 indicates strong support, and $BF > 10$
306	are considered very strong support <sup>27</sup> . We inferred the <i>r</i> values of <i>BMR</i> and $T_b$ with the
307	phylogenetic variable-rates regression models that best fit the data for our samples of
308	mammals and birds (Table S7 and S8). We also estimated the $r$ values for $T_a$ after
309	accounting for the effect of latitude of species distribution (Table S11) and, consequently,
310	we accounted for the geographic variation of $T_a$ across extant species distributions. We
311	used BayesTraits v3.0 <sup>28</sup> to detect the magnitude and location of $r$ in a Bayesian Markov
312	chain Monte Carlo (MCMC) reversible-jump framework, which generates a posterior
313	distribution of trees with scaled branches lengths according to the rate of evolution. There
314	is no limit or prior expectation in the number of the <i>r</i> branch-scalars, <i>r</i> numbers vary
315	from zero (no branch is scaled) to $n$ , where $n$ is the number of branches in the
316	phylogenetic tree. Regarding the values of each r parameter, we used a gamma prior,
317	with $\alpha = 1.1$ and $\beta$ parameter rescaled in order to get the median of the distribution equal
318	to one. With this setting, the numbers of rate increases and decreases proposed is
319	balanced <sup>13</sup> . We ran 50,000,000 iterations sampling every 25,000 to ensure chain
320	convergence and independence in model parameters in $BMR$ and $T_b$ analyses. We
321	discarded the first 25,000 iterations as burn in. For the $T_a$ analysis in mammals we ran
322	200,000,000 iterations sampling every 100,000, and we discarded the first 100,000
323	iterations as burn in. For $T_a$ analysis in birds we ran 400,000,000 iterations discarding the
324	first 100,000,000 as burn in, and we sampled every 200,000. Regression coefficients
325	were judged as significant according to a calculated $p_{\text{MCMC}}$ value for each posterior of
326	regression coefficients: where < 5% of samples in the posterior distribution crossed zero,
327	this indicates that the coefficient is significantly different from zero.

329	Testing the relationship between the branch-wise rates of evolution. We first
330	estimated the consensus branch-scaled tree for $BMR$ and $T_b$ from the posterior sample of
331	branch-scaled trees obtained with the phylogenetic variable-rates regression model. The
332	consensus branch-scaled tree was generated by using the median $r$ from the posterior
333	distribution. We evaluated the correlation between the $r$ values for <i>BMR</i> and $T_b$ using a
334	Bayesian GLS regression in BayesTraits v3.0. The same analyses were conducted to
335	evaluate the correlation between <i>branch-wise rates</i> for <i>BMR</i> and $T_a$ . We used a uniform
336	prior for the $\beta$ (slope coefficient) ranging from -100 to 100. We ran 50,000,000 iterations
337	sampling every 25,000 to ensure chain convergence and independence in model
338	parameters. Significance of regression coefficients were determined as above.
339	
340	<b>Detecting trends.</b> We evaluated the direction of change in <i>BMR</i> , $T_b$ , and $T_a$ across all
340 341	<b>Detecting trends.</b> We evaluated the direction of change in <i>BMR</i> , $T_b$ , and $T_a$ across all mammals and birds using the <i>path-wise rates</i> of these variables (Table S15 and S16).
341	mammals and birds using the <i>path-wise rates</i> of these variables (Table S15 and S16).
341 342	mammals and birds using the <i>path-wise rates</i> of these variables (Table S15 and S16). <i>Path-wise rate</i> is the sum of all the <i>branch-wise rates</i> along the path of a species, which
<ul><li>341</li><li>342</li><li>343</li></ul>	mammals and birds using the <i>path-wise rates</i> of these variables (Table S15 and S16). <i>Path-wise rate</i> is the sum of all the <i>branch-wise rates</i> along the path of a species, which lead from the root (the MRCA) to the tips of the tree, and it accounts for the total changes
<ul><li>341</li><li>342</li><li>343</li><li>344</li></ul>	mammals and birds using the <i>path-wise rates</i> of these variables (Table S15 and S16). <i>Path-wise rate</i> is the sum of all the <i>branch-wise rates</i> along the path of a species, which lead from the root (the MRCA) to the tips of the tree, and it accounts for the total changes the species has experienced during its evolution <sup>15</sup> . If high <i>path-wise rates</i> have
<ul> <li>341</li> <li>342</li> <li>343</li> <li>344</li> <li>345</li> </ul>	mammals and birds using the <i>path-wise rates</i> of these variables (Table S15 and S16). <i>Path-wise rate</i> is the sum of all the <i>branch-wise rates</i> along the path of a species, which lead from the root (the MRCA) to the tips of the tree, and it accounts for the total changes the species has experienced during its evolution <sup>15</sup> . If high <i>path-wise rates</i> have disproportionately been associated with trait increases or decreases, we expect to find that
<ul> <li>341</li> <li>342</li> <li>343</li> <li>344</li> <li>345</li> <li>346</li> </ul>	mammals and birds using the <i>path-wise rates</i> of these variables (Table S15 and S16). <i>Path-wise rate</i> is the sum of all the <i>branch-wise rates</i> along the path of a species, which lead from the root (the MRCA) to the tips of the tree, and it accounts for the total changes the species has experienced during its evolution <sup>15</sup> . If high <i>path-wise rates</i> have disproportionately been associated with trait increases or decreases, we expect to find that species with greater <i>path-wise rates</i> will have high or low trait values in the present. For
<ul> <li>341</li> <li>342</li> <li>343</li> <li>344</li> <li>345</li> <li>346</li> <li>347</li> </ul>	mammals and birds using the <i>path-wise rates</i> of these variables (Table S15 and S16). <i>Path-wise rate</i> is the sum of all the <i>branch-wise rates</i> along the path of a species, which lead from the root (the MRCA) to the tips of the tree, and it accounts for the total changes the species has experienced during its evolution <sup>15</sup> . If high <i>path-wise rates</i> have disproportionately been associated with trait increases or decreases, we expect to find that species with greater <i>path-wise rates</i> will have high or low trait values in the present. For instance, if ancestral mammals experienced progressively colder environmental

351	BayesTraits 3.0 to evaluate the relationship between <i>BMR</i> , $T_b$ , $T_a$ and their <i>path-wise</i>
352	rates (Table S15 and S16). We used a uniform prior for the $\beta$ (slope coefficients) ranging
353	from -100 to 100 to allow all possible values to be equally likely. Finally, we ran
354	50,000,000 iterations sampling every 25,000 to ensure chain convergence and
355	independence in model parameters. Significance of regression slopes were determined as
356	above.
357	
358	End notes.
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372	
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375	Data availability statement. Correspondence and request for materials should be
376	addressed to J.A-LL. (jorgeavariall@gmail.com) or C.V. (c.d.venditti@reading.ac.uk).
377	
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