

Heritability and developmental plasticity of growth in an oviparous lizard

Fonti Kar¹, Shinichi Nakagawa^{1,2}, Daniel W.A. Noble^{1,3}

¹ School of Biological Earth and Environmental Sciences, Ecology and Evolution Research Centre, University of New South Wales, Sydney, NSW, Australia

² Diabetes and Metabolism Division, Garvan Institute of Medical Research, 384 Victoria Street, Darlinghurst, Sydney, NSW 2010, Australia

³ Division of Ecology and Evolution, Research School of Biology, The Australian National University, Canberra, ACT, Australia

Corresponding author: Fonti Kar

Correspondence email: fonti.kar@gmail.com

Abstract

Selective processes act on phenotypic variation yet the evolutionary potential of any given trait relies on underlying heritable variation. Developmental plasticity is an important source of phenotypic variation, but it can also promote changes in heritability by modifying environmental sources of variability. Here, we quantified the influence of developmental temperature on an important fitness trait, growth, in delicate skinks (*Lampropholis delicata*). We partitioned the total phenotypic variance using an animal model fitted with a genomic relatedness matrix. We measured mass growth for 262 individuals ($n_{\text{hot}} = 126$, $n_{\text{cold}} = 136$) over 16 months ($n_{\text{observations}} = 3,002$); estimating heritability and maternal effects over time from animals experiencing two thermal developmental environments. Our results show that lizards reared in cold developmental temperatures had a higher initial mass compared to lizards that were reared in hot developmental temperatures. However, developmental temperature did not impact the rate of growth. On average, additive genetic variance, maternal effects and heritability were higher in the ‘hot’ developmental temperature treatment. Interestingly, heritability increased with age, whereas maternal effects decreased upon hatching but increased again at a later age. Our work suggests that evolutionary potential of growth is complex, age dependent and not overtly affected by extremes in natural nest temperatures.

Keywords

Body mass, growth rate, additive genetic variance, incubation temperature, maternal effects, temperature-size rule, cryptic genetic variation

37 Introduction

38 Developmental plasticity plays a key role in generating phenotypic variation (Noble et
39 al 2018; Ghalambor et al., 2007; West-Eberhard, 2003). The complex interplay between an
40 individual's genotype, and the developmental environment in which that genotype finds
41 itself, means that a range of different phenotypes can arise (Monaghan, 2008; West-Eberhard,
42 2003). Phenotypic changes resulting from distinct early life experiences can have persistent
43 effects on individual fitness (Monaghan, 2008; Noble et al., 2018). Changes induced by
44 developmental environments may result in a better match between the adult phenotype and
45 the subsequent selective environment. However in some cases, maladaptive phenotypes can
46 arise if there is a mismatch between later-life environments and those experienced early in
47 development (Beaman et al., 2016; Ghalambor et al., 2007). Regardless, phenotypic plasticity
48 represents a promising immediate solution for threatened populations by allowing them to
49 better track adaptive optima and persist (Beldade et al., 2011; Noble et al., 2019; West-
50 Eberhard, 2003). Understanding the consequences of developmental environments on
51 phenotypes and fitness is therefore critical to predict how populations will survive in stressful
52 conditions (Botero et al., 2015; Reed et al., 2010).

53
54 A population's capacity to evolve depends not only on the strength of selection but
55 also on the underlying standing genetic variation (Lynch & Walsh, 1998). It has long been
56 recognised that selection and genetic variation change across environments (Falconer &
57 Mackay, 1996). As such, a great deal of effort has been put towards understanding the
58 circumstances under which genetic variation may change with the environment and the
59 magnitude of those changes (Charmantier & Garant, 2005; Fischer et al., 2020; Hoffmann &
60 Merilä, 1999; Noble et al., 2019; Rowiński & Rogell, 2017; Wood & Brodie, 2015). Genetic
61 variance in novel environments may decrease as a result of stronger selection that erodes
62 genetic variation (Hoffman & Parsons, 1991; Hoffmann & Merilä, 1999). In contrast, novel
63 environments might also increase genetic variance when mutation rates are higher or
64 buffering mechanisms breakdown triggering a release of 'cryptic genetic variation' in
65 stressful conditions (Paaby & Rockman, 2014). Low cross-environment genetic correlations
66 or condition-dependence of gene expression can also affect the amount of genetic variance in
67 different environments (Charmantier & Garant, 2005; Coltman et al., 2001). Environmental
68 dependence of genetic variance implies that under the same selection pressure, the speed of
69 evolutionary change will be expected to change making it difficult to predict genetic
70 adaptation.

71
72 Comparative studies have shown that the influence of environmental stress on genetic
73 variance during development is not straightforward (Charmantier & Garant, 2005; Hoffmann
74 & Merilä, 1999; Rowiński & Rogell, 2017). In lab studies, elevated developmental stress has
75 been shown to increase the heritability of morphological traits (Hoffmann & Merilä, 1999),
76 whereas wild, non-domestic populations tend to have higher heritability in favourable
77 environments (Charmantier & Garant, 2005). Lack of consensus may be related to increased
78 environmental heterogeneity in wild populations, making them more difficult to compare
79 with lab studies. It has been suggested that responses to different developmental stressors
80 (e.g. heat shock vs. starvation) may be associated with disparate patterns of gene expression
81 making broad comparisons more variable (Charmantier & Garant, 2005; Dahlgaard &
82 Hoffmann, 2000). Importantly, environmental comparisons of heritability have been
83 criticised as the ratio nature of its calculations can mask changes in the relative contributions
84 of non-genetic and genetic variance (Rowiński & Rogell, 2017). For example, a meta-
85 analysis found that heritability of life history traits which has been argued to be more
86 important to fitness, did not change between control and stressful conditions (Rowiński &

87 Rogell, 2017). The same pattern was observed for morphological traits (Fischer et al., 2020).
88 Upon closer inspection, both additive genetic and environmental variance of life history traits
89 increased under stressful conditions whereas the opposite was true for morphological traits
90 (Rowiński & Rogell, 2017). The dynamics of both genetic and non-genetic sources of
91 variation under different developmental environments can thus influence the evolutionary
92 potential of fitness related traits.

93

94 Body size is fundamental to fitness and is both heritable and environmentally
95 responsive (Noordwijk et al., 1988; Stillwell & Fox, 2009). Developmental environments,
96 such as temperature and nutritional stress can drive substantial variation in body size, largely
97 through shifts in how organisms grow (Eyck et al., 2019; Noble et al., 2018). Maternal
98 investment in offspring are also important sources of body size variation (Noble et al., 2014;
99 Wilson & Réale, 2006). Variation among mothers in egg investment, nest site selection or
100 timing of birth (Mitchell et al., 2018; Shine & Harlow, 1996; Uller & Olsson, 2010) are
101 expected to contribute the most to offspring body size early in development (Mousseau &
102 Fox, 1998). However, these effects have shown to decline with age as maternal investment
103 subside (Krist, 2010; Wilson, Kruuk, et al., 2005). Additionally, environmental factors such
104 as shared habitats or long-term seasonal effects can also account for a substantial proportion
105 of variability in body size (Kruuk, 2004). For example, permanent environmental effects that
106 varied across years explained 26% – 35% of body size variation in bighorn sheep (Réale et
107 al., 1999). Similarly, 56% of variation in body mass was attributed to nest boxes shared
108 among siblings in blue tit chicks (Charmantier et al., 2004). As such, the various sources that
109 influence body size variation (genetic, environmental, maternal) are predicted to vary across
110 ontogeny and temporal approach is therefore needed in order to evaluate when evolutionary
111 potential of body size is greatest.

112

113 Here we investigated the impact of developmental temperature on body size (mass)
114 and growth in an oviparous skink (*Lampropholis delicata*) – two traits that are critically
115 important to fitness. We also test how developmental environments affect evolutionary
116 potential in these traits. Growth trajectories ($n_{\text{observations}} = 3,002$) for lizards that hatched from
117 two incubation treatments ($n_{\text{hot}} = 126$, $n_{\text{cold}} = 136$), were measured over the first 16 months of
118 life (nearly half their life). Using 8,433 single nucleotide polymorphic (SNP) markers, we
119 derived a genomic relatedness matrix to estimate quantitative genetic parameters. Using these
120 data, we address two key questions: 1) How does developmental temperature affect the rate
121 and shape of growth trajectories (initial mass, growth rate and curvature of growth
122 trajectory)? and 2) How does developmental temperature affect genetic and non-genetic
123 variance across age? According to the ‘temperature-size rule’, we expect lizards experiencing
124 cold developmental temperatures to have larger initial masses and slower growth rates –
125 possibly resulting in lizards reaching sexual maturity at a later age compared to lizards
126 experiencing hot developmental temperatures (Angilletta Jr et al., 2017). In addition, we
127 predicted greater amount of genetic variance under higher developmental temperatures, after
128 controlling for non-genetic sources of variance. We expected maternal effects and permanent
129 environment effects to manifest early in development and dissipate over time.

130 **Materials and Methods**

131 *Lizard collection and husbandry*

132 From 2015 – 2017, we established a breeding colony of adult *L. delicata* ($n_{\text{females}} =$
133 144, $n_{\text{males}} = 50$) using wild individuals collected across five sites throughout the Sydney
134 region between August and September 2015. Using a half-sib breeding design, we paired
135 three females with a single male in opaque plastic enclosures measuring 35cm × 25cm

136 × 15cm (L × W × H). Enclosures were kept under UV lights (12L:12D) in a temperature-
137 controlled room set to 24°C. Lizards were given access to a heat lamp that elevated
138 temperatures to between 28-32 °C. Each enclosure was lined with newspaper and lizards had
139 constant access to water. Tree bark was used as refuge. Adult lizards were fed medium sized
140 crickets *ad libitum* (*Acheta domestica*) dusted with calcium powder and multi-vitamin every
141 two days. From the beginning of the egg laying season (October of each year), we replaced
142 newspaper lining with garden potting mix and placed an opaque plastic box (12 cm × 17.5
143 cm × 4.3 cm) containing moistened vermiculite in each enclosure for females to oviposit
144 their eggs. During this time, enclosures were sprayed with water every second day to
145 maintain a relatively humid environment. From October to November, egg boxes were
146 checked every day. Tail tissue samples (~1 mm) were taken from adults that were from
147 enclosures producing eggs for DNA extraction (see below). All tissues were stored in 70%
148 ethanol. Animal collection was approved by the New South Wales National Parks and
149 Wildlife Service (SL101549) and all procedures were approved by the Macquarie University
150 Ethics committee (ARA 2015/015) and University of New South Wales Animal Care and
151 Ethics committee (ACEC 15/51A).

152

153 *Developmental Temperature Manipulations*

154 Eggs were collected between October to March, over two reproductive seasons from
155 2016 and 2017. As soon as eggs were found, they were weighed using a digital scale to the
156 nearest 0.01g (Ohaus Scout SKX123). We also measured egg length (distance between the
157 furthest points along the longest axis of the egg) and egg width (distance between the widest
158 points along the axis perpendicular to the longest axis of the egg) using digital callipers to the
159 nearest 0.01mm. Following measurements, each egg was placed in a plastic cup (80ml)
160 containing three grams of vermiculite and four grams of water. Each cup was then covered
161 using cling wrap and secured using an elastic band. We used a split-clutch design where eggs
162 from single clutch were pseudo-randomly assigned to one of two developmental temperature
163 treatments. We used two incubators to precisely control the temperature of eggs (LabWit,
164 ZXSD-R1090). The ‘hot’ treatment was exposed to a mean temperature of 29°C whereas the
165 ‘cold’ treatment was exposed to a mean temperature of 23°C. Both incubators fluctuated +/-
166 3°C over a 24-hour period around these mean temperatures to simulate natural nest site
167 temperature variability. These treatments represent the temperature extremes of natural nest
168 sites for *L. delicata* (Cheetham et al., 2011). Egg cups were rotated within each incubator
169 weekly to avoid uneven heat circulation within incubators. Incubators were also checked
170 daily for hatchlings.

171

172 *Quantifying Growth Rate*

173 Newly emerged hatchlings were weighed to the nearest 0.01g and a small tail tip
174 clipping (~2mm) was taken for genetic analyses. Ventral photographs were taken for digital
175 measurement (Nikon Coolpix A900). For the first two months, photographs of hatchlings
176 were taken approximately every 14 days. After which, hatchlings were photographed at
177 approximately a 35-day interval. From six months onwards, we manually measured hatchling
178 SVL using a clear ruler to the nearest ~0.5mm. We also recorded the mass of the individual
179 each time photographs or SVL measurements were taken. Growth measurements continued
180 until we had approximately 16 measures per individual (mean = 11.5, SD = 4.71). By the end
181 of the study, the mean age for hot incubated lizards was 335.82 (range: 0 – 711) and for cold
182 incubated lizards it was 384.8 (range: 0 – 707) which is approximately 40 – 50% of their total
183 lifespan (Chapple et al., 2014). From the photographs, we extracted snout-vent-length (SVL;
184 from tip of snout to the beginning of the cloaca opening) using ImageJ software (Rueden et al.,
185 2017). For the first initial nine months, hatchlings were housed individually in opaque plastic

186 enclosures (32.3cm x 18.5cm x 6cm) lined with newspaper. Hatchlings were fed the same
187 number of crickets every second day and had constant access to a tree bark refuge and water.
188 Hatchling enclosures were placed in a temperature control room under the same conditions as
189 described above for the adult colony. For logistical reasons, at approximately nine months,
190 hatchlings were housed in groups of five in opaque bins with the same measurements as the
191 adult enclosures. We pseudo-randomised individuals to each shared enclosure while
192 maintaining a similar number of individuals from each treatment.

193 *Genomic Relatedness Matrix*

194 We derived a genomic relatedness matrix (GRM) using single nucleotide
195 polymorphism (SNP) genotypes for all 262 offspring with growth data (132 putative parents;
196 $n_{\text{females}} = 69$, $n_{\text{males}} = 63$). While our half-sib breeding design allowed us to assign parentage to
197 derive a pedigree, high levels of sperm storage and low levels of multiple paternity (94% of
198 females had been sired by a single male) meant our pedigree had low resolution to effectively
199 estimate additive genetic variation. Recent studies have shown that GRM derived from SNPs
200 have low error rates (<0.3%) and are able to reconstruct pedigree relationships in much finer
201 detail when at least 200 SNP loci are used (Béréños et al., 2014; Huisman, 2017). Moreover,
202 both relatedness and heritability values estimated from a GRM have been shown to be very
203 similar to those inferred using a pedigree (Béréños et al., 2014; Huisman, 2017). Single
204 nucleotide polymorphism libraries were designed and animals genotyped using DArTseq™ (
205 Diversity Arrays Technology) methods. For more details on DNA extraction and SNP
206 genotyping see ESM.

207 Prior to deriving our GRM, we filtered our SNPs using the R package *dartR* (Gruber et
208 al., 2018). We filtered loci based on various metrics in the following order: 1) read depth (8 –
209 40); reproducibility (> 0.996); call rate by loci (> 0.97) and then by individual (> 0.80);
210 monomorphic loci; minor allele frequencies (> 0.02); Hamming Distance among loci (> 0.25)
211 and Hardy Weinberg Equilibrium. This clean-up process resulted in a dataset of 8,438 loci with
212 an average call rate of 98.5% (see ESM and provided code). Using these 8,438 loci we derived
213 a GRM, which describes the proportion of the genome that is identical by descent (VanRaden,
214 2008). We calculated a GRM for all hatchlings using the *snpReady* R package (Granato et al.,
215 2018) following methods described by VanRaden, 2008:

$$216 \quad \quad \quad 217 \quad \quad \quad \text{GRM} = \frac{ZZ'}{2 \sum p_i(1 - p_i)}$$

218 where Z is the centered squared matrix of SNP genotypes of all individuals. This is calculated
219 from a matrix of where heterozygote SNP genotypes (AT) were coded as 0, homozygote
220 genotypes for the SNP allele (AA) were coded as 1 and homozygotes for the original allele
221 (TT) were coded as -1. p_i is the frequency of the second locus at locus position i . The
222 denominator scales the GRM matrix so that the values approximate a relatedness matrix
223 derived from a pedigree. The GRM was then inverted for modelling fitting (see ESM and
224 provided code).
225

226 **Statistical Analyses**

227 All analyses were performed using *R* (Core Team, 2013). We checked the data for
228 potential input errors using histograms, scatterplots and Cleveland plots. We fitted Bayesian
229 linear mixed effects models (LMM) in *brms* with interfaces with Stan (Bürkner, 2017;
230 Gelman et al., 2015). Mass was log-transformed, and age was z-transformed. For all models
231 we used noninformative priors with 4000 iterations with a burn in of 1500, sampling from the

232 posterior distribution every fifth iteration. We ensured proper mixing by inspecting trace
 233 plots and checked that scale reduction factors were less than 1.01. We report posterior means
 234 and 95% credible intervals for all parameters throughout.

235 *Impact of Developmental Temperature on Additive Genetic Variance and Maternal Effects*
 236 *Across Age*

237 First, we tested whether developmental temperature influenced the overall heritability
 238 of mass and the relative contributions of variance irrespective of age. For each treatment
 239 group, we fitted intercepts only in the fixed effects with random intercepts for additive
 240 genetic variance (G), maternal effects (M) and permanent environmental effects (PE) as we
 241 had repeated measures of the same individuals (Wilson et al., 2010). The model also
 242 estimated residual variance (R). We included our GRM to estimate additive genetic variation.
 243 Overall, Heritability (h^2) of mass using this intercept (I) model was calculated as:

244
$$h^2 = \frac{G_I}{(G_I + M_I + PE_I + R_I)}$$

245 To then test how G , M and h^2 change across age, we used model selection to
 246 determine the most appropriate random effects structure for our data as we had no *a priori*
 247 knowledge of what (or how) variance components change with age (Wilson & Réale, 2006).
 248 We fitted seven models with varying complexity in their random effects and compared their
 249 Watanabe–Akaike Information Criterion (WAIC) values (Table S1). We fitted random
 250 intercepts and random slopes by including either a linear age term or both linear and
 251 quadratic age terms to partition variance across age. Two models were equally supported, the
 252 first included a random linear and quadratic slope for G and M and PE . (Model 3 - Table S1)
 253 and the second included a random linear and quadratic slope for G and M , respectively, and a
 254 random intercept for PE (Model 7 – Table S1). To avoid overfitting, we selected the more
 255 parsimonious model and used this random effect structure for the remaining analyses unless
 256 stated otherwise.

257 Residual variance may be conflated with estimates of other variance components if it
 258 changes over time (heterogenous variance) and is not properly accounted for. We therefore
 259 explicitly modelled residual variance to verify if this was the case and compared homogenous
 260 and heterogenous residual variance models using WAIC. We fitted two models, both of
 261 which had the same fixed and random effects structure as Model 7 described above. The first
 262 model had homogenous residual variance whereas in the second model we modelled residual
 263 variance with a linear slope thereby allowing it to vary with age. The model with
 264 heterogenous variance was best supported (Table S2), we therefore modelled heterogenous
 265 variance in all subsequent models unless stated otherwise.

266 To test for treatment differences in variance components, we subset data for each
 267 treatment group and fitted an intercept-only model with our best supported random effect
 268 structure (Model 7) and heterogenous residual variance. We estimated a genetic variance-
 269 covariance matrix for each treatment (G), where the diagonal elements represent the additive
 270 genetic variances for the intercept (G_I), slope (G_S) and the quadratic (G_C) across age. The off-
 271 diagonal elements are the additive genetic covariances between the growth curve parameters,
 272 for example, $Cov_{I,C}$ is the additive genetic variance between the intercept and the quadratic
 273 slope.

274
$$G = \begin{bmatrix} G_I & Cov_{I,S} & Cov_{I,C} \\ Cov_{I,S} & G_S & Cov_{S,C} \\ Cov_{I,C} & Cov_{S,C} & G_C \end{bmatrix}$$

275 Similarly, the variance-covariance matrix for dams (M) can be decomposed in the same
 276 manner as G .

$$277 \quad M = \begin{bmatrix} M_I & Cov_{I,S} & Cov_{I,C} \\ Cov_{I,S} & M_S & Cov_{S,C} \\ Cov_{I,C} & Cov_{S,C} & M_C \end{bmatrix}$$

278 For each treatment group, we then calculated additive genetic variance at a given age G_x
 279 using the random slope terms and their covariances following (Schielzeth & Nakagawa,
 280 2020):

$$281 \quad G_x = G_I + (x^2 \cdot G_S) + (x^4 \cdot G_C) + (2x \cdot Cov_{I,S}) + (2x^2 \cdot Cov_{I,C}) + (2x^3 \cdot Cov_{S,C})$$

282 where x is a specific age. Age-specific maternal effect M_x was calculated using the same
 283 formula but with the relevant variance components from M . Age-specific heritability, h_x^2 , is
 284 thus a ratio of all variance components at a given age x . The proportion of variance explained
 285 by maternal effects (m^2) is calculated in the same manner.

$$286 \quad h_x^2 = \frac{G_x}{(G_x + M_x + PE_I + R_I)}$$

287 As the mean body mass increases over time, the variance may also increase
 288 concurrently due to scale effects and potentially bias estimates of quantitative genetic
 289 parameters (Wilson, Kruuk, et al., 2005). We therefore calculated coefficients of variation
 290 (CV) across age for each variance component by dividing variance by the predicted mean
 291 mass at a given age. Interpretations using CV estimates did not change our overall
 292 conclusions for additive genetic variance or maternal effects, we therefore present the raw
 293 estimates of each variance component below (See ESM).
 294

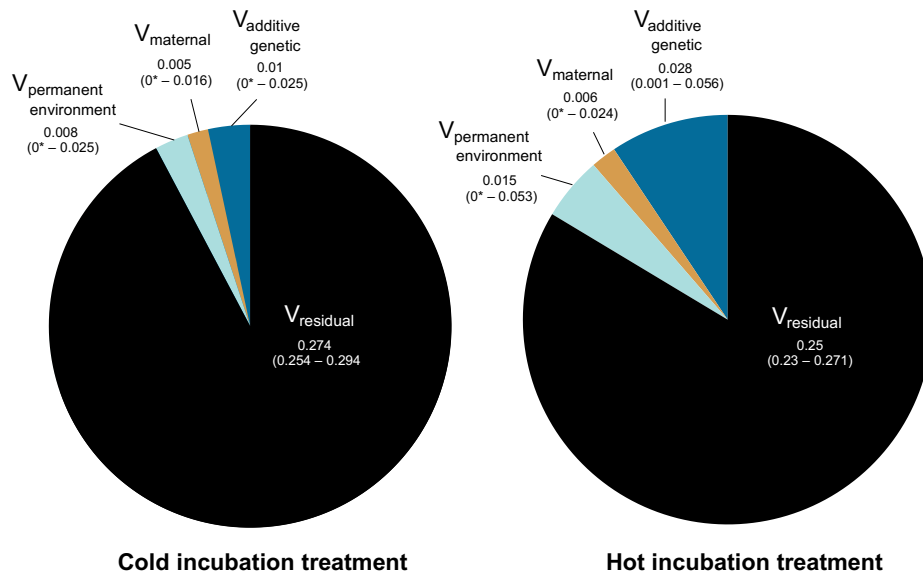
295 *The Influence of Developmental Temperature on Growth Trajectories*

296 To test how developmental temperatures affect average growth trajectories, we also
 297 fitted three models that varied in their fixed effect structure to determine how developmental
 298 temperatures affect: 1) initial mass (intercept of curve), 2) linear rate of growth (linear slope)
 299 and 3) curvature of the growth trajectory (quadratic term). We also wanted to test for
 300 treatment differences in age at which lizards reach their maximum mass by solving for the
 301 maxima of quadratic regression equation. We fit mass as the response accounting for the
 302 same random effects described above. The first model included the main effect of
 303 developmental temperature and the linear and quadratic term for age (Table S2). The other
 304 two models differed in their interaction terms between developmental temperature with age
 305 and age² (Table 2, S3). We then compared WAIC values to select the best model for our data
 306 that explained changes in mass across age between the two developmental temperature
 307 treatments (Table 1).
 308

309 **Results**

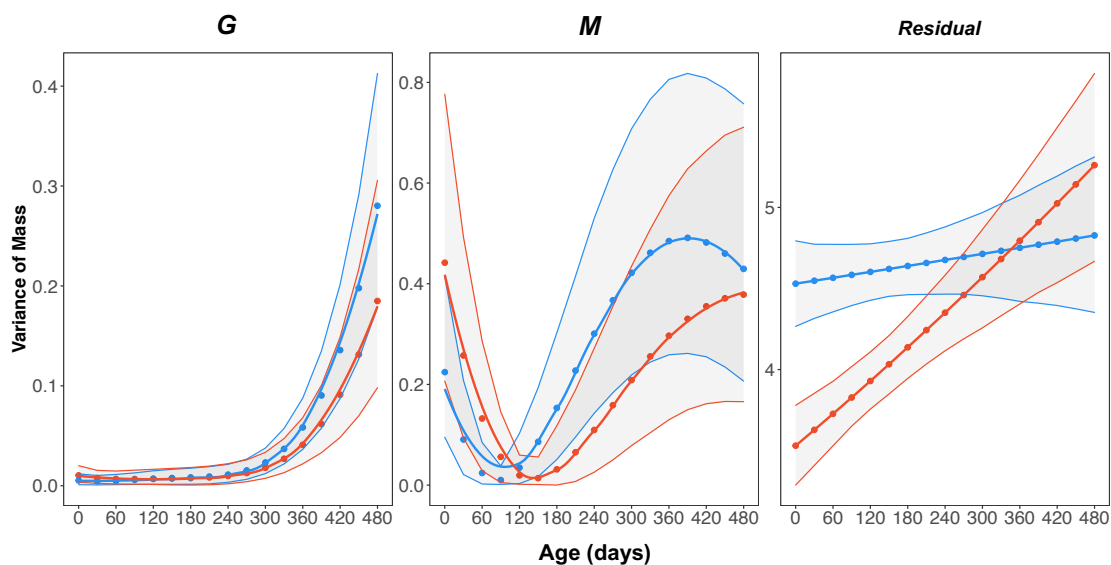
310 Over two years, we collected 3,002 observations of mass data for a total of 261
 311 individuals ($n_{hot} = 125$, $n_{cold} = 136$). On average, the incubation period for the ‘hot’
 312 treatment was 29.36 days (SD = 2.17, range = 15 - 49) days and 48.48 days (SD = 4.18, range
 313 = 25 - 56) for the ‘cold’ treatment.

314 Overall, additive genetic variance, permanent environmental variance and heritability
 315 (h^2) of growth appears to be higher in the hot developmental temperature treatment (Fig. 1).
 316 However, there were no significant differences among treatment groups (Table S3).



317 **Figure 1** Pie charts depicting the overall relative contributions of mass variance for the hot
 318 ($n_{\text{lizards}} = 126$) and cold ($n_{\text{lizards}} = 136$) developmental treatment group irrespective of age.
 319 Point estimates and 95% credible intervals are presented in Table S3. There were no
 320 significant differences in variance components between developmental temperature
 321 treatments. * in indicates very small values that were above 0. The influence of developmental
 322 temperature on genetic and non-genetic variance across age

323 Treatment groups did not differ in how the relative contributions of G and M changed with
 324 age as their 95% credible intervals overlapped (Fig. 2). Additive genetic variance remained
 325 relatively low and constant upon emergence until approximately nine months of age, after
 326 which it increased rapidly (Fig. 2). Maternal effects decreased sharply upon hatching and
 327 dropped to the minimum at approximately six months before it increased again (Fig. 2).
 328 There were some differences among developmental treatments in how residual variance



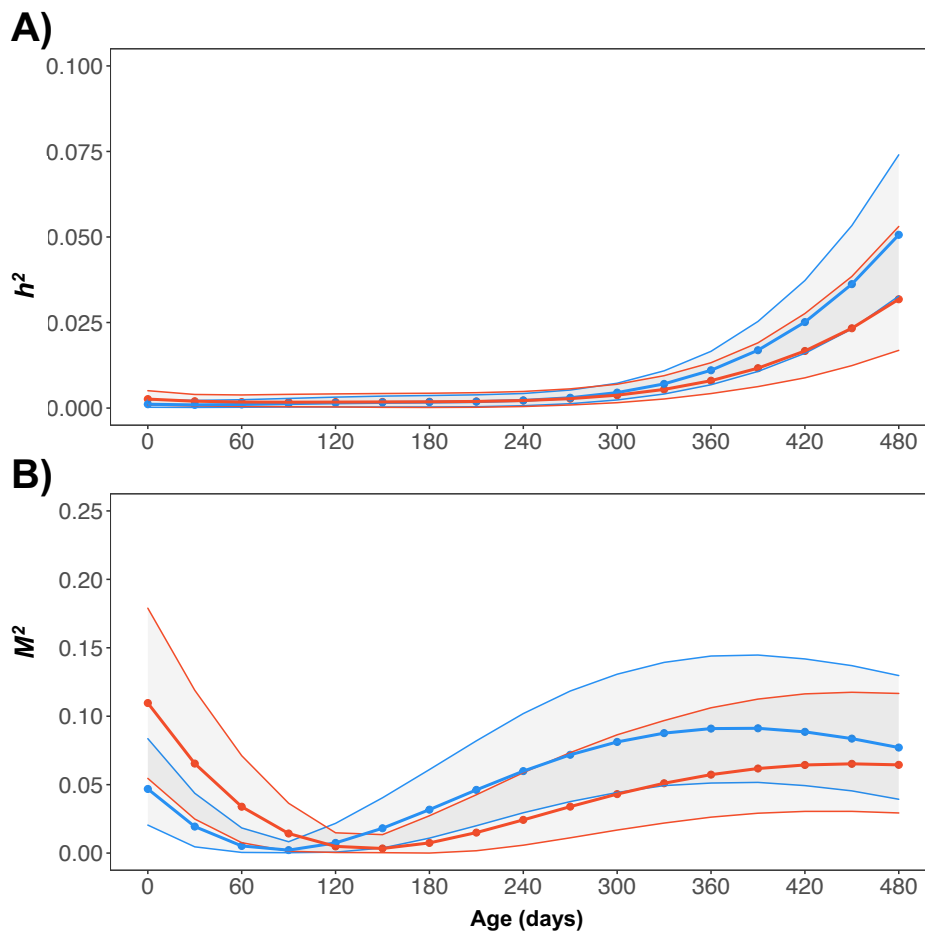
329 changed with age (Fig. 2C). Residual variance in cold incubated lizards had a much higher
330 intercept compared to hot incubated lizard however their residual variance converged by
331 eight months of age (Fig. 2).

332 **Figure. 2** Scatterplot showing how additive genetic variance (G), maternal effects (M),
333 residual variance changed with age for the hot developmental treatment ($n_{\text{lizards}} = 125$, red)
334 and the cold developmental treatment ($n = 136$, blue). Points represent posterior means, thin
335 lines represent the 95% credible intervals, thick lines represent the mean for each treatment
336 group. Note that permanent environmental effects were treated as constant across age.
337 $V_{\text{permanent environment}}$ for the hot treatment group was 0.0047 [0.00017 – 0.0096], $V_{\text{permanent}}$
338 environment for the cold treatment group was 0.0047 [0.00065 – 0.0085].

339 We investigated whether increases in average mass over time affected variance
340 estimates due to scaling effects between the mean and variance. However, we found that the
341 CV of G and M followed the same pattern as the raw variance estimates suggesting that
342 changes in variance were not the result of increasing mean body mass with age (Fig. S1).

343 After accounting for heterogenous residual variance, we found no treatment
344 differences in heritability or the proportion of variance explained by maternal effects (M^2)
345 (Fig. 3). Heritability was very low for the first year of growth in *L. delicata* and only began
346 increasing at one year of age (Fig. 3). As predicted M^2 decreased soon after hatching,
347 however it increased slightly again from six months of age (Fig. 3). The G and M matrices for
348 each treatment group are presented in Table S4-S5.

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376 **Figure 3** Heritability (h^2 , **A**) and the proportion of total variance explained by maternal effect
 377 variance (M^2 , **B**) across age (days) for the hot developmental treatment ($n_{\text{lizards}} = 125$, red) and
 378 the cold developmental treatment ($n_{\text{lizards}} = 136$, blue). Points represent estimates generated
 379 from the posterior distribution of the variance-covariance matrix, thin lines represent the 95%
 380 credible intervals, thick lines represent the mean for each treatment group.

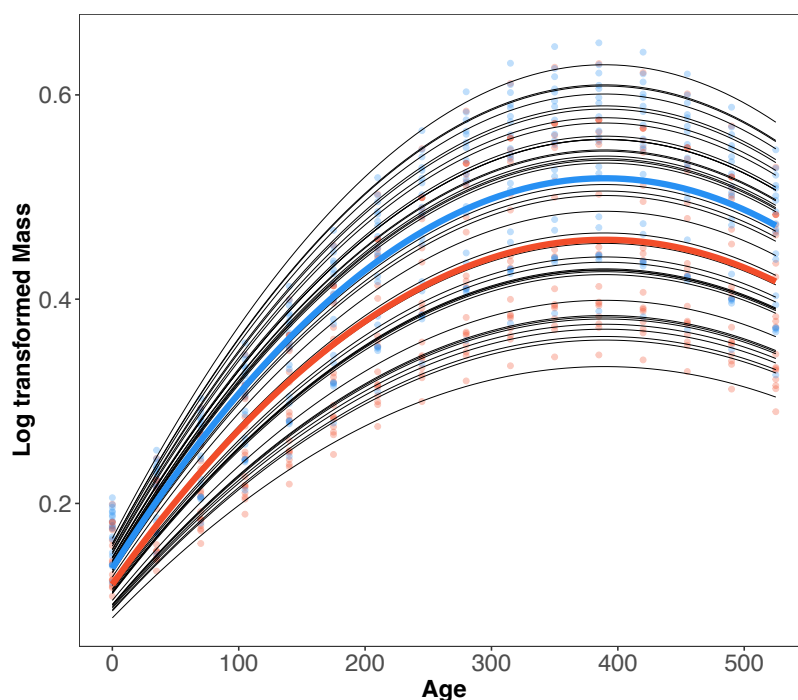
381 *Developmental plasticity in growth trajectories in response to temperature*

382 While the model containing a full interaction between treatment and linear and
 383 quadratic age was best supported, the improvement in WAIC value was marginal (Table 1).
 384 Moreover, the linear growth rate (Age) and curvature of the growth trajectory (Age²) did not
 385 differ significantly between the two developmental temperature treatments in any of the
 386 models containing interactions (Table S7 - S9). Irrespective of treatment, lizard mass
 387 increased by 1.65 g for every 1 SD unit increase in age.

388 **Table 1** Comparisons of WAIC values of four models ($n_{\text{obs}} = 2926$) with different
 389 combinations of treatment interactions with age parameters. $\Delta ELPD$ represents the difference
 390 in expected log predicted density. Age measured in days was z-transformed (mean = 361.34,
 391 SD = 185.16)

Formula of Fixed Effects	WAIC	$\Delta ELPD$	Std. Error $\Delta ELPD$
Treatment + Age + Age ² + Treatment × Age + Treatment × Age ²	-3301	0	0
Treatment + Age + Age ² + Treatment × Age	-3295	-0.62	1.182
Treatment + Age + Age ² + Treatment × Age ²	-3300	-2.798	1.375
Treatment + Age + Age ²	-3292	-4.452	1.563

392
 393 Developmental temperature did, however, influenced hatching mass (Table 1, Fig. 3).
 394 Lizards from the ‘cold’ treatment were on average 0.030 g (0.018g – 0.041g) heavier
 395 compared to lizards from the ‘hot’ treatment (Table. 2). Larger initial masses meant that
 396 lizards from the ‘cold’ treatment reached their maximum mass slightly earlier (382.97 days,
 397 95% CI: 358.84 – 409.78) compared to lizards from ‘hot’ treatment (413.04 days, 95% CI:
 398 379.70 – 452.34). G and M matrices from this model, along with other variance components,
 399 are presented in Table S6.



400

401 **Figure 3** Model predictions of log-transformed mass over age from the two developmental
 402 temperatures. We randomly subset 40 lizards (20 from each treatment) to plot their individual
 403 growth curves. Points represent mean estimates for each lizard from the hot developmental
 404 treatment (hot) and the cold developmental treatment (blue). Thick lines represent average
 405 growth curve for each treatment. Faint grey lines are each individual's growth curve. Model
 406 predictions were generated from the full model where interaction terms between treatment
 407 and both the linear component and quadratic component were included

408

409 **Table 2** Coefficient estimates from full model testing the effects of developmental treatment
 410 on mass and how mass changes with age. Bolded estimates are significantly different from
 411 zero. * indicates that value is above zero prior to rounding. $n_{\text{obs}} = 2926$. Age measured in
 412 days was z-transformed (mean = 361.34, SD = 185.16). G and M matrices for this model is
 413 presented in Table S6.

Parameter	Estimate	Lower	Upper
Intercept	-0.991	-1.01	-0.971
Treatment	-0.083	-0.114	-0.05
Age	0.5	0.476	0.526
Age ²	-0.196	-0.216	-0.178
Treatment × Age	0.008	-0.021	0.037
Treatment × Age ²	0.022	-0.007	0.052

414

415 Discussion

416 Early development at hot temperatures resulted in smaller body sizes compared to
 417 development at cold temperatures. Growth trajectories, however, were not significantly
 418 impacted by early thermal environments – lizards from both temperatures grew at the same
 419 rate despite cold animals remaining larger throughout life. Marginalising over age, we found

420 that developmental temperature did not impact the relative contributions of additive genetic,
421 maternal, permanent environment or residual variance. The environmental component of the
422 phenotype (residual variance) explained most of the variability in body mass. Congruently,
423 heritability of mass was generally low across ontogeny, increasing at one year of age. As we
424 predicted, maternal effects on offspring mass declined in the first few months, presumably
425 because maternal non-genetic contributions were less influential on mass over time.
426 Unexpectedly, maternal effects increased again at approximately six months possibly from
427 maternal genetic factors affecting mass. Upon hatching, the residual variance component of
428 body mass was much higher in lizards that were reared at cold incubation temperatures,
429 suggesting that aspects of development environment played a bigger role in determining their
430 hatching mass.

431 *Thermal developmental plasticity in growth*

432
433 In ectotherms, temperature plays a pervasive role in phenotypic development (Eyck et al.,
434 2019; Noble et al., 2018; O’Dea et al., 2019; While et al., 2018). Contrary to other reptile
435 studies, we did not show that growth rate differed between developmental temperatures.
436 Some researchers reported increases in growth at higher incubation temperatures (Elphick &
437 Shine, 1999; Hare et al., 2004; Verdú-Ricoy et al., 2014), while have others found either the
438 opposite result or no differences at all (Andrews et al., 2000; Goodman, 2008). The
439 directionality of change is highly variable, even among studies of the same species (e.g.
440 *Bassiana dupreyi*, Elphick & Shine, 1998, 1999; Flatt et al., 2001; Telemeco et al., 2010).
441 Lack of generality may be related to how growth is statistically modelled. Very few studies
442 account for individual variation in hatching mass or growth trajectories. Indeed, if we did not
443 account for among individual variance in our models, significant treatment differences in
444 growth can be detected (Table S10). We emphasise the importance of partitioning
445 confounding sources of variance such as individual or clutch effects as they can misconstrue
446 conclusions about developmental impacts on late life phenotypes. Moreover, future studies
447 should make use of all repeated measures of mass instead of averaging across individuals as
448 the former approach not only increases statistical power but also provide more accurate
449 estimates of growth.

450
451 Consistent with other squamates, we found that the cold incubation treatment group attained
452 higher hatching mass compared to their hot counterparts (Dayananda et al., 2016; Downes &
453 Shine, 1999; Flatt et al., 2001; Goodman et al., 2013). These results support the temperature-
454 size-rule whereby organisms reared in cold temperatures tend to have larger body sizes
455 (Angilletta Jr et al., 2017). Larger hatching size can be achieved through prolonged
456 development at cooler temperatures during embryonic stages (Forster & Hirst, 2012). It is
457 well known that cold developmental temperatures result in longer incubation periods in many
458 reptiles (Booth, 2006; Dayananda et al., 2016; Downes & Shine, 1999; Elphick & Shine,
459 1998; R. M. Goodman, 2008). Longer developmental time may allow embryos to assimilate
460 yolk nutrients more efficiently thus increasing mass at hatching (Storm & Angilletta, 2007).
461 Indeed, turtle embryos exposed to high temperatures have enhanced mitochondrial
462 metabolism and metabolic enzymic activity which constrains developmental time and
463 reduced overall hatching size (Ji et al., 2003; Sun et al., 2015). Thermal plasticity in
464 embryonic development may be adaptive for lizards born late in the season when nest
465 temperatures are generally colder (Warner & Shine, 2008; While et al., 2015). Indeed, female
466 *L. delicata* have an extended oviposition period (September to February in our population)
467 and nest temperatures during this time can be highly variable in the wild (Cheetham et al.,
468 2011). Heavier weight at emergence could mean that hatchlings are in better condition to

469 compete with lizards that hatched earlier or have sufficient body reserves to survive harsher
470 conditions in more seasonal environments (Downes & Shine, 1999; Gifford et al., 2017;
471 Qualls & Shine, 2000). Understanding how body mass affects survival will be necessary to
472 elucidate the adaptive potential of developmentally plastic responses in the wild.
473

474 *Thermal developmental environments and the evolutionary potential of body mass*

475
476 Adaptive evolutionary responses depend not only on the amount of selection operating on a
477 trait but on also its underlying additive genetic variance (Falconer, 1952; Ghilambor et al.,
478 2007; Hoffmann & Merilä, 1999). Stressful developmental environments are hypothesized to
479 lead to the release of ‘cryptic’ genetic variation (Fischer et al., 2020; Noble et al., 2019;
480 Rowiński & Rogell, 2017; Wood & Brodie, 2015), possibly increasing the evolutionary
481 potential of a given trait. Higher genetic variation, combined with stronger selection may
482 facilitate rapid evolutionary responses that may allow populations to adapt to novel
483 environments (Hoffmann & Merilä, 1999; Falconer and Mackay 1996). Contrary to these
484 hypotheses, we found no statistical differences in additive genetic variance for mass between
485 our developmental temperature treatments. In fact, heritability for mass was overall quite low
486 echoing heritability values for mass in various animal systems [e.g., bighorn sheep – 0.03 to
487 0.31 (Réale et al., 1999), macaques – 0.39 (Kimock et al., 2019) lizards – 0 to 0.54 – (Martins
488 et al., 2019; Noble et al., 2014)]. It should be noted that decoupling additive genetic variances
489 from other non-genetic variance such as maternal effects requires considerable paternal links
490 in the study design and pedigree (Kruuk, 2004). Indeed, when this variance partitioning is
491 done accordingly, heritability estimates are often low (e.g., Noble et al. 2014). In the case of
492 our study, we found relatively low levels of multiple paternity (<1% of clutches were sired by
493 multiple fathers), as such the number of half-sibs were generally low which may have
494 affected our genomic relatedness matrix and estimates of quantitative genetic parameters.
495

496 Lack of differences in genetic variation between developmental temperature
497 environments support findings from recent meta-analyses. Fisher et al. (2020) assessed the
498 degree to which stressful thermal environments result in the release of genetic variation. They
499 found that these effects manifested in only a third of the studied cases – in mainly clonal
500 organisms (Fischer et al., 2020). Furthermore, of the 25 cases where genetic variance
501 changed across thermal environments there was no consistent direction (i.e., 11 increased and
502 14 decreased under thermal stress). Noble et al. (2019) also showed that the release of
503 ‘cryptic’ genetic variation depends on the study design – studies not able to partition out non-
504 genetic sources of variation supported a release of genetic variation whereas studies that did
505 showed the opposite pattern. As a caveat, defining an environment as stressful or novel is a
506 difficult task which requires detailed knowledge of a given species’ past environmental
507 exposure – information that is often unknown (Roelofs et al., 2010). While our incubation
508 temperatures were selected based on temperature extremes of naturally occurring *L. delicata*
509 nests (Cheetham et al., 2011), it is nonetheless possible they were not ‘stressful’ from an
510 evolutionary perspective. Indeed, egg mortality did not differ across incubation treatments
511 which suggests that lizards from both treatments experienced a similar level of thermal stress
512 as embryos (the estimate of treatment difference: 0.80 [-0.04 -1.73]). Furthermore, treatment
513 differences may be harder to detect under realistic fluctuating temperature regimes. As such,
514 lizards were not exposed to extreme temperatures over extended periods which might be
515 more important in orchestrating changes in genetic variation (Bonamour et al., 2019).
516 Overall, our results suggest that the thermal extremes experienced by natural nest sites do not
517 modify the evolutionary potential of mass. However this should be interpreted with caution

518 as estimates of quantitative parameters from laboratory studies can differ from wild
519 populations (Sgrò & Hoffmann, 2004; Weigensberg & Roff, 1996).
520

521 *Ontogenetic changes in genetic and non-genetic contributions to body mass*

522
523 Genetic contributions to body size are expected to vary throughout ontogeny (Lynch &
524 Walsh, 1998). Selection pressures on body size are likely to increase at critical life stages,
525 such as at birth or at sexual maturation, thereby reducing genetic variance at certain ages
526 (Rollinson & Rowe, 2015). On the contrary, we found that additive genetic variance of mass
527 was very low upon hatching but slowly increased by the end of the first year. This result
528 parallels those seen in big horn sheep (Réale et al., 1999), soay sheep (Wilson et al., 2007)
529 and ladybird beetles (Dmitriew et al., 2010). While the underlying cause of this pattern is not
530 well established, it coincided with changes in the social environment (shared housing). This
531 suggests that perhaps competition for resources (basking sites or food) may orchestrate
532 changes in genetic variation (Dmitriew et al., 2010; Hoffmann & Merilä, 1999).
533 Alternatively, the gradual increase in additive genetic variance may be related to initial
534 genotypic changes underpinning sexual maturation (~14 months) as *L.delicata* are sexually
535 dimorphic in various morphological traits including body size (Chapple et al., 2014).
536 Nonetheless, ontogenetic variation in genetic variance implies that potential rates of
537 evolution varies with age (Houle, 1998), however this depends on non-genetic sources of
538 variance as well.

539
540 Maternal non-genetic contributions to offspring body size are expected to be highest during
541 early life stages and decline as offspring mature, particularly in precocial species (Cheverud,
542 1984; Wilson, Kruuk, et al., 2005). In accordance with other studies, maternal effects did in
543 fact decline after hatching (Dmitriew et al., 2010; Lindholm et al., 2006; Pick et al., 2016;
544 Wilson, Coltman, et al., 2005; Wilson, Kruuk, et al., 2005). Maternal investment, such as
545 investment in clutch number or egg quality, has been shown to influence hatching size in
546 lizards (Brown & Shine, 2009; Noble et al., 2014; Warner & Lovern, 2014), however, as
547 predicted, these effects dissipated post-hatching (Pick et al., 2016; Réale et al., 1999).
548 Interestingly, maternal contributions increased at a later age and remained moderately low for
549 the remainder of the study. The cause of resurgence in maternal effect variance is unclear,
550 however this pattern may indicate other maternally inherited components such as maternal
551 genetic effects (e.g., mitochondrial genetic variation) that promote variation in body size
552 (Pick et al., 2016). Indeed, variation in mitochondrial function has been linked to an
553 individual's metabolic rate and growth – explaining as much as ~50% of the variation in food
554 intake and growth (Salin et al., 2016, 2019). Therefore, it is likely an important driver of
555 body size variability. Similar to additive genetic variance, resurgence of maternal effects also
556 cooccurred with changes in the shared environment (housing conditions), suggesting that
557 maternal effects on offspring body size is likely to be environmentally driven.

558
559 Traits under strong selection are expected to show low evolutionary potential as selection acts
560 to remove genetic variation. While low evolutionary potential is at least in part due to
561 reduced levels of additive genetic variance, it is also a result of larger proportions of
562 environmental variance. In our study, the environmental component of the phenotype
563 accounted for over 80% of variation in body mass which is in line with values reported in
564 great tits (53 – 74%) and soay sheep (70 – 96%) (Noordwijk et al., 1988; Wilson et al., 2007).
565 Interestingly, cool developmental temperatures increased the amount of environmental
566 variance attributed to body mass at an early age. What mechanisms are comprised in this

567 environmental component? Variation in developmental period between developmental
568 temperatures may explain these differences. In many ectotherms, developmental time
569 exhibits a nonlinear reaction norm with temperature (Marshall et al., 2020; Noble et al.,
570 2018). This means that developmental time decelerates with temperature following an
571 negative exponential function. As a result, hot incubated lizards are more constrained in their
572 development time compared to lizards that were reared a cooler temperature. In actual fact,
573 the cold developmental temperature treatment had much greater variance in incubation
574 duration. With a longer incubation period, embryos can maximise the yolk resources left by
575 their mothers which can vary considerably within clutch (Wallace et al., 2007). Our results
576 suggest that thermodynamic effects of development time can give rise greater environmental
577 heterogeneity in hatching mass and may affect potential for evolution at early life stages.

578 **Conclusion**

579 Our work illustrates the pervasive role of developmental temperature on phenotypic
580 variation. The impact of developmental temperature on body mass manifested early and
581 persisted through life (Monaghan, 2008). This has profound implications as developmentally
582 induced variation in body mass may drive life history differences within populations and alter
583 their vulnerability to environmental change (Botero et al., 2015; Marshall et al., 2020; Reed
584 et al., 2010). In contrast, genetic variance of body mass was robust to thermal extremes
585 experienced by natural nests and suggests that the potential to genetically adapt to warming
586 climate may be limited. However, more stressful incubation temperatures are needed to
587 elucidate the capacity for this species to reveal new genetic material for selection to act on.
588 Non-genetic sources of variance were responsible for most of the variability in body mass
589 and their dynamics with age means that effectiveness of evolution is everchanging.
590 Understanding the complexities of adaptive evolution in response to climate change may
591 require intensive long-term studies in wild populations.
592

593 **Author contributions**

594 FK, DN and SN conceived the study, FK and DN collected and analysed the data, FK wrote
595 the first draft, FK, DN and SN edited the manuscript.

596 **Data accessibility**

597 Datasets and code used to generate results of this study is accessible via Open Science
598 Framework (<https://bit.ly/2Uy72id>)
599

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604

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