

1 **Temperature-driven selection on metabolic traits increases the**
2 **strength of an algal-grazer interaction in naturally warmed**
3 **streams**

4 **Running head:** Linking metabolism and trophic interactions
5

6 **Authors**

7 C. -Elisa Schaum^{1,4*}, Richard French-Constant², Chris Lowe^{1,2}, Jón S. Ólafsson³, Daniel
8 Padfield¹, Student Research Team¹ & Gabriel Yvon-Durocher^{1*}
9

10 **Author affiliations**

11 1 Environment and Sustainability Institute, University of Exeter, Penryn, Cornwall TR10 9EZ, UK

12 2 Centre for Ecology and Conservation, College of Life and Environmental Sciences, University of Exeter,
13 Penryn, Cornwall, TR10 9FE, U.K.

14 3. Marine and Freshwater Research Institute, Árleyni 22, 112 Reykjavik, Iceland.

15 4. Institute for Hydrobiology and Fisheries, Centre for Earth System Research and Sustainability, Section
16 Biological Oceanography, University of Hamburg, Hamburg, 22767, Germany

17 * Corresponding authors: elisa.schaum@uni-hamburg.de, g.yvon-durocher@exeter.ac.uk,
18

19 **Student Research Team:**

20 Yasmin Ashton, Romina Botoli, Peter Coles, Joe Crisp, Emma Dwan, Stella Enoch-Pledger,
21 Briony Ffello, Kate Freegard, Ceri Haines, Matthew Holland, Luke Lear, Emma
22 Lokuciejewski, Heather McPhee, Toby Newport, Liisa Pahk, Sienna Somers, Caspar
23 Swindells, Maria Wild, Ethan Wrigglesworth
24
25

26 **Author Contributions:**

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44 **ABSTRACT**

45 Trophic interactions are important determinants of the structure and functioning of
46 ecosystems. Because the metabolism and consumption rates of ectotherms increase sharply
47 with temperature, there are major concerns that global warming will increase the strength of
48 trophic interactions, destabilizing food webs, and altering ecosystem structure and function.
49 We used geothermally warmed streams that span an 11°C temperature gradient to investigate
50 the interplay between temperature-driven selection on traits related to metabolism and
51 resource acquisition, and the interaction strength between the keystone gastropod grazer,
52 *Radix balthica*, and a common algal resource. Populations from a warm stream (~28°C) had
53 higher maximal metabolic rates and optimal temperatures than their counterparts from a cold
54 stream (~17°C). We found that metabolic rates of the population originating from the warmer
55 stream were higher across all measurement temperatures. A reciprocal transplant experiment
56 demonstrated that the interaction strengths between the grazer and its algal resource were
57 highest for both populations when transplanted into the warm stream. In line with the thermal
58 dependence of respiration, interaction strengths involving grazers from the warm stream were
59 always higher than those with grazers from the cold stream. These results imply that
60 increases in metabolism and resource consumption mediated by the direct, thermodynamic
61 effects of higher temperatures on physiological rates are not mitigated by metabolic
62 compensation in the long term, and suggest that warming could increase the strength of
63 algal–grazer interactions with likely knock-on effects for the biodiversity and productivity of
64 aquatic ecosystems.

65

66 **Keywords:** Consumer-resource interactions, global warming, metabolism, thermal
67 adaptation, interaction strength

68

69 INTRODUCTION

70 The strength of consumer-resource interactions (e.g. the effect of a consumer on the
71 population density of its prey) plays a critical role in shaping the stability of food webs (May,
72 1973; McCann, Hastings, & Huxel, 1998; Otto, Rall, & Brose, 2007; Paine, 1980). Grazing is
73 an important class of consumer–resource interaction, determining the flux of energy and
74 materials from autotrophs to heterotrophs. There are currently major concerns that global
75 warming will increase the impact of grazers on algal or plant communities because the
76 ingestion and respiration rates of heterotrophs tend to increase more rapidly with rising
77 temperatures than rates of photosynthesis and growth in autotrophs (Gilbert et al., 2014;
78 O’Connor, 2009; West & Post, 2016). Stronger interactions have the potential to destabilize
79 food webs and consequently, warming induced increases in interaction strengths could have
80 fundamental implications for ecosystem structure and function. For example, elevated
81 grazing rates in aquatic ecosystems, driven by the mismatch in thermal sensitivity between
82 autotrophs and heterotrophs, are hypothesized as a key driver of projected declines in aquatic
83 primary production over the 21st century in models of ocean biogeochemistry (Laufkötter et
84 al., 2015).

85 The effects of temperature on metabolic rates and traits associated with consumer–resource
86 interactions (e.g. attack rates, handling times) often follow characteristic unimodal thermal
87 response curves, where rates increase exponentially to an optimum and decline rapidly
88 thereafter (Dell, Pawar, & Savage, 2011, 2014; Englund, Oehlund, Hein, & Diehl, 2011;
89 Gilbert et al., 2014; Rall et al., 2012). Integrating thermal responses for metabolism and
90 interaction traits with dynamical models of consumer–resource interactions offers a
91 promising framework for predicting food web responses to global warming (Binzer, Guill,
92 Rall, & Brose, 2015; Shurin, Clasen, Greig, Kratina, & Thompson, 2012; Vasseur &
93 McCann, 2005). However, thermal response curves are often flexible, and can shift when

94 organisms are exposed to novel thermal environments, both via phenotypic plasticity, where
95 organisms change phenotypic characteristics rapidly and with no underlying heritable genetic
96 change (West-Eberhard, 2003), and adaptive evolution, where organisms respond to changes
97 in the environment through heritable genetic change over many generations, resulting in
98 better adapted phenotypes (Angilletta, Wilson, Navas, & James, 2003; Deutsch et al., 2008;
99 Kingsolver & Huey, 2008; Kingsolver, Ragland, & Shlichta, 2004). Consequently, plasticity
100 and evolution have the potential to modulate the effects of rising temperatures on the strength
101 of species interactions (Sentis, Morisson, & Boukal, 2015). For example, if metabolic rates
102 are down-regulated after long-term exposure to higher temperatures (Addo-Bediako, Chown,
103 & Gaston, 2000), then compensatory metabolic responses to warming (e.g. Padfield et al.,
104 2017) could mitigate predicted increases in consumer–resource interaction strength. How
105 these long-term responses to warming affect rates of metabolism and in turn, the strength of
106 consumer– resource interactions, are largely unknown, limiting our ability to predict how
107 trophic interactions will change in response to warming in the long term.

108 There is evidence from studies across naturally occurring thermal gradients over large spatial
109 scales, that local thermal adaptation can play an important role in shaping the strength of
110 species in shaping the stability of food webs (May, 1973; McCann, Hastings, & Huxel, 1998;
111 Otto, Rall, & Brose, 2007; Paine, 1980). Grazing is an important class of consumer–resource
112 interaction, determining the flux of energy and materials from autotrophs to heterotrophs.

113 There are currently major concerns that global warming will increase the impact of grazers
114 on algal or plant communities because the ingestion and respiration rates of heterotrophs tend
115 to increase more rapidly with rising temperatures than rates of photosynthesis and growth in
116 autotrophs (Gilbert et al., 2014; O’Connor, 2009; West & Post, 2016). Stronger interactions
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119 function. For example, elevated grazing rates in aquatic ecosystems, driven by the mismatch
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140 are down-regulated after long-term exposure to higher temperatures (Addo-Bediako, Chown,
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144 consumer– resource interactions, are largely unknown, limiting our ability to predict how
145 trophic interactions will change in response to warming in the long term.

146 There is evidence from studies across naturally occurring thermal gradients over large spatial
147 scales, that local thermal adaptation can play an important role in shaping the strength of
148 species interactions (Barton, 2011; De Block, Pauwels, Van Den Broeck, De Meester, &
149 Stoks, 2012). While these studies provide important insights into how consumer–resource
150 interactions are shaped by evolution across thermal gradients (Fukami & Wardle, 2005), their
151 usefulness for understanding responses to rapid climate warming might be limited because
152 other factors, such as day length, light intensity and precipitation, tend to be confounded with
153 temperature along such broad scale spatial gradients. Furthermore, the timescales over which
154 local adaptation has occurred in such broad scale studies could be much longer than the rapid
155 evolutionary change required to keep pace with climate warming (Hoffmann & Sgro , 2011;
156 Loarie et al., 2009). Here, we investigate how temperature-driven selection on traits that
157 determine the thermal responses of metabolism and resource acquisition affect the strength of
158 a keystone grazing interaction (the gastropod *Radix balthica*, which grazes algal biofilms in
159 streams) in naturally warmed Icelandic geothermal streams spanning a gradient of 11°C.

160 Critically, temperature is the main abiotic factor that varies among streams in the catchment
161 and is not correlated with pH, conductivity or inorganic nutrient concentrations (see Table 1,
162 also Padfield et al., 2017). These streams are thought to have been subject to geothermal
163 heating for at least the last century (O’Gorman et al., 2012). This system therefore provides
164 the opportunity to investigate how long-term differences in temperature between otherwise
165 similar sites shape the expression of metabolic traits and the subsequent impact of any
166 temperature-driven selection on species interactions in a natural system. Specifically, we test
167 the hypothesis that long-term differences in temperature drive selection for metabolic traits
168 that dampen the direct effects of warming on metabolic rates and the strength of consumer–

169 resource interactions. We predict that (i) snails from warm streams will have down-regulated
170 rates of metabolism when normalized to a reference temperature, and (ii) consequently, the
171 effects of warming on algal–grazer interaction strengths involving snails from warm streams
172 will be attenuated relative to expectations based on the direct effects of rising temperature
173 alone.

174

175

176 **MATERIALS AND METHODS**

177

178 **Study site**

179 The streams are located North of the Hveragerði valley, in the south east of the Hengil high
180 temperature geothermal field, Iceland (N64° 0' 2.944" W21° 11' 17.451") and consist of a
181 catchment of 11 streams spanning a temperature gradient of approximately 20 °C (see Figure
182 1 and Figure S1). Two streams, stream 5 (17.5 °C ± 4.5 °C, hereafter ‘cold stream’) and
183 stream 11A (28.3 °C ± 1.3 °C, hereafter ‘warm stream’, see Table 1 for a comparison of other
184 chemical and physical parameters), were chosen for experiments due to their close proximity
185 to each other, the large temperature differential and similar abundances of the keystone
186 grazer, *Radix balthica*. The grazer plays an important functional role in geothermal stream
187 ecosystems, where grazer biomass as well as grazing rates are strongly influenced by
188 temperature (OGorman *et al.*, 2012). The two streams are similar in all other measured
189 physical and chemical characteristics, including stream velocity at the sampling location, but
190 differ in average temperature by 11 °C (see Table 1), and hence present an opportunity to
191 investigate how the effects long-term differences in temperature shape consumer-resource
192 interactions. The experiment was carried out from 26.05.2016 to 02.06. 2016, with a pilot
193 study testing the general experimental set-up conducted from 01.06.2015 to 07.06. 2015.

194

195 **Grazer metabolism**

196 To quantify whether the different thermal regimes in the two adjacent streams resulted in
 197 divergence in metabolic traits of mature *R. balthica* (based on mass and shell length) we
 198 measured the acute responses of respiration to a broad gradient in temperature. For the
 199 metabolic measurements, we collected 33 individuals of similar mass and length (average
 200 mass of snails from cold stream $0.087 \text{ g} \pm 0.014\text{g}$, average mass of snails from warm stream
 201 0.098 ± 0.016 , both ± 1 SEM, ANOVA: $F_{1,61}$, $P = 2.15$, see Table S1 and also compare
 202 Figure S2 for mass of another 205 randomly collected snails) from each stream, which were
 203 cleaned from any algal debris to avoid carry-over of a food source into the tank or subsequent
 204 respiratory measurements on the oxygen electrode. The snails were kept for 24hrs in aerated
 205 tanks at the average stream temperature of origin and in the absence of a food source to
 206 minimise any potential effects of differences in food quantity or quality between streams.
 207 Respiration was quantified as the rate of oxygen consumption in a Clark-Type oxygen
 208 electrode, measured between 4 – 44 °C in 4 °C increments (11 temperatures in total) within a
 209 day. At each temperature, respiration was measured for 3 individuals, and a different set of
 210 individuals was measured at each temperature (i.e. each animal was only subjected to a single
 211 assay). Individuals were allowed 15 minutes at the assay temperature prior to the
 212 measurements. The subsequent thermal responses of respiration were quantified using a
 213 modification of the Sharpe-Schoolfield equation (see (Schoolfield *et al.*, 1981) and (Sharpe &
 214 DeMichele, 1977)for the original equation):

$$215 \ln(b(T)) = E_a \left(\frac{1}{kT_c} - \frac{1}{kT} \right) + \ln(b(T_c)) + \alpha \ln(M_i) - \ln \left(1 + e^{E_h \left(\frac{1}{kT_h} - \frac{1}{kT} \right)} \right) \quad (1)$$

216 where $b(T)$, is the *per capita* metabolic rate ($\mu\text{mol O}_2 \text{ L}^{-1} \text{ h}^{-1}$) at temperature T in Kelvin (K),
 217 k is Boltzmann's constant ($8.62 \times 10^{-5} \text{ eV K}^{-1}$), E_a is an apparent activation energy (in eV) for
 218 the metabolic process, $\ln(b(T_c))$ is the rate of metabolism normalised to an arbitrary
 219 reference temperature, $T_c = 18 \text{ °C}$, where no low or high temperature inactivation is
 220 experienced. M_i is the mass (g) of an individual i , α is the allometric scaling exponent that

221 characterises the power-law relation of mass and metabolic rate (Brown *et al.*, 2004). E_h
222 characterizes temperature-induced inactivation of enzyme kinetics above T_h where half the
223 enzymes are rendered non-functional. Differentiating equation (1) and solving for the global
224 maxima yields an expression for the optimum temperature

$$225 \quad T_{opt} = \frac{E_h T_h}{E_h + k T_h \ln\left(\frac{E_h}{E_a} - 1\right)} \quad (2)$$

226 Equation (1) differs from the Sharpe-Schoolfield equation (Sharpe & DeMichele, 1977;
227 Schoolfield *et al.*, 1981) in a number of ways. First, we account for the power law relation
228 between body mass and metabolic rate, M^α (Brown *et al.*, 2004). Second, we exclude
229 parameters from Eq. (1) used to characterize low-temperature inactivation due to insufficient
230 data to quantify this phenomenon in our analysis. Third, rather than characterize temperature
231 effects below T_{opt} using the Eyring (1935) relation, $\left(\frac{T}{T_c}\right) e^{E_a\left(\frac{1}{kT_c} - \frac{1}{kT}\right)}$, we instead use the
232 simpler Boltzmann factor, $e^{E_a\left(\frac{1}{kT_c} - \frac{1}{kT}\right)}$. This simplification enables an explicit solution for T_{opt}
233 (Eq. 2) and facilitates more direct comparison with previous work on the temperature
234 dependence of metabolism using metabolic theory e.g. (Allen *et al.*, 2005, Gillooly, 2001;
235 Brown *et al.*, 2004; Van M Savage *et al.*, 2015).

236 The parameters, $\ln b(T_c)$, α , E_a , E_h , T_h , and T_{opt} , in Eqs. (1) & (2) represent traits
237 characterising the metabolic thermal response that we expect to be under selection in *R.*
238 *balthica* inhabiting the hot and cold streams. We tested for differences in each of the
239 parameters between the populations of *R. balthica* by fitting the respiration data to Eq. (1)
240 using generalised non-linear least squares regression (within the ‘gnls’ function in the ‘nlme’
241 package for R, package version 3.1-128) and including ‘origin’ as a two level factor (i.e.
242 ‘cold’ and ‘warm’ stream). We tested for differences between populations for each parameter
243 by sequentially removing the effect of ‘origin’ on each parameter and comparing the Akaike
244 information criterion for small sample sizes (AICc) for all possible models (see Table S1 and

245 Table S3) using the ‘aictab’ and ‘modavg’ functions from the AICcmodavg package
246 (package version 2.1-0). The model chosen for further exploration was that with the lowest
247 (AICc) value. Model averaging was carried out when models fell within 2 AICc units of each
248 other, and the conditional averages of the parameters were used for curve fitting and
249 interpretation (see also Table 2). The relative importance of the fixed factors in the averaged
250 model was determined using the sum of their relative weights.

251

252 **Reciprocal transplant experiment**

253 We carried out a reciprocal transplant experiment to determine how long-term differences in
254 temperature and the resultant impacts on metabolic traits affect the strength of algal-grazer
255 interactions. We achieved this by placing adult snails from each population in microcosms
256 consisting of a tissue culture flask on which diatom biofilms had been established. There was
257 no significant difference in size or mass of the snails chosen for the reciprocal transplant
258 experiment. Snails from the cold stream weighed approximately $0.83 \pm 0.08\text{g}$, and snails
259 from the warm stream, $0.89\text{g} \pm 0.07\text{g}$ (no significant difference, see also Table S4, ANOVA
260 $F_{1,65} = 0.81$, $P = 0.76$). Diatoms of the genera, *Acnantes*, *Nitzschia*, *Navicula*, and
261 *Gomphonema* are common in streams across the Hengill volcanic area (Gudmundsdottir *et al.*
262 2013) and were ordered from culture collections (Culture collection of algae and protozoa
263 and Sciento) and grown in the laboratory in mixed assemblages to yield common resource for
264 testing the effects of temperature and local adaptation on grazing . The diatom assemblages
265 were inoculated into Corning plastic translucent flasks (maximum volume 1L) with 20 mL
266 COMBO medium (Kilham *et al.*, 1998), and brought to a salinity of 5-10 (equivalent to
267 approximately 5-10 g salts/kg water) to match the slightly elevated salinity and conductivity
268 found in these thermal stream environments (Gudmundsdottir *et al.* 2013). The flasks were
269 turned onto their sides to allow for a larger area of biofilm growth on the base ($\sim 60\text{ cm}^2$ in

270 total per flask) and the algal communities were left to grow for 14 days prior to the
271 experiment. After 14 days, all flasks had substantial biofilm development on the base and
272 were used as microcosms for the *in situ* reciprocal transplant experiment. Analysis of control
273 flasks (no grazer) showed that growth of the diatom lawn *per se* did not differ significantly
274 for flasks placed in hot or cold streams (Figure S3, one-way ANOVA $F_{1,10} = 1.28$, $P = 0.26$).
275 Thus, any changes to the biofilm biomass in the experiment can be attributed to the per capita
276 effects of the grazer.

277 The experiment consisted of 3 treatments (each with 6 replicate microcosms placed in
278 each of the 2 streams): (i) a control microcosm in which a biofilm was present and no *R.*
279 *balthica* were added, (ii) an ‘origin’ treatment in which *R. balthica* that were resident in the
280 stream were added to microcosms, and (iii) a ‘transplanted’ treatment in which *R. balthica*
281 that were from the adjacent stream were added to microcosms. *R. balthica* individuals used
282 for the reciprocal transfer experiment were collected from the 2 streams prior to the
283 experiment and were starved for 48h in the laboratory in aerated tanks at the average
284 temperature of the stream of origin. Microcosms were assembled by adding 3 snails of
285 similar body dimensions (0.35 ± 0.03 g of *R. balthica* mass reported as blotted fresh weight
286 throughout) and 100 mL of 0.4 μm filtered water from the stream in which the microcosm
287 was to be placed. This resulted in a grazer density of 5 individuals m^{-2} , which was
288 comparable to the average *in situ* density in the streams (see Figure S4, no significant
289 difference in *in situ* density between the two-streams: one-way ANOVA: $F_{1,66}$, $P = 0.54$).
290 This design was preferred to a set-up with each microcosm holding a single grazer, which
291 attempt to exclude the effects of mutual interference on feeding behaviour e.g. (Skalski &
292 Gilliam, 2001; Rall *et al.*, 2010; Lang *et al.*, 2011; Vucic-Pestic *et al.*, 2011), because (i) the
293 experimental densities are representative of natural conditions; and (ii) the consumption rates
294 of a single individual were insufficient to detect a significant change in algal biomass. The

295 microcosms were submerged in each stream facing downstream, so that the stream water
296 controlled the ambient temperature of the microcosms but did not fill the flasks further with
297 water or organic matter (see Figure 1 for conceptual graphic). The snails were left to graze
298 for 48 hours. We observed no grazer mortality over the experimental period. Sampling of
299 another 206 snails from each stream confirmed no significant difference in average snail size
300 in the cold and warm stream at this time of the year (see Figure S2, one-way ANOVA: $F_{1,408}$
301 = 0.15, $P = 0.7$).

302

303

304

305

306 **Interaction strength**

307 At the end of the experiment, algal biomass in each of the microcosms was quantified via
308 methanol chlorophyll extraction modified from (Holm-Hansen & Riemann, 1978). Here, the
309 walls of the microcosms were scrubbed until all biofilm particles were in suspension. The
310 solution was filtered onto a 0.4µm GF/F filter, which was then ground in methanol for 5
311 minutes. The samples were centrifuged at 3500 rpm for 15 minutes and the absorbance of the
312 supernatant was measured at 632nm, 665nm, and 750nm. Total chlorophyll content in µg
313 mL⁻¹ was then calculated as described in Holm-Hansen & Riemann (1978). The *per capita*
314 interaction strength in each microcosm was then estimated by calculating the dynamic index
315 (DI, see also (Berlow *et al.*, 2004) for a technically similar set-up): Note that in Berlow *et al.*
316 2004, N is for the grazed, and D for the ungrazed habitats. We chose to invert this ratio for a
317 more intuitive comparison of metabolic rate and grazing data.

318

$$319 \quad DI = \frac{\ln\left(\frac{N}{D}\right)}{Yt} \quad (3)$$

320 where DI is the dynamic index (g Chl gC⁻¹ h⁻¹), *N* is total chlorophyll (sum of Chl *a* + Chl *c*)
321 content of control, *D* total chlorophyll in the grazed microcosm, *Y* is the grazer biomass (g
322 C), and *t* is time in hours. Snail blotted wet weight was converted to carbon mass (in grams)
323 using conversion factors that assume dry weight to be 7.5% of the blotted wet weight
324 (Ricciardi & Bourget, 1998) and a carbon content of 22% dry weight (Burgmer *et al.*, 2010).

325 We carried out two analyses using the data from the reciprocal transplant experiment.
326 The first analysis, used a generalised linear model (GLM), with ‘interaction strength’ as the
327 response variable and ‘origin’ (‘cold’ or ‘warm’ stream) and ‘transplant temperature’ (17.5
328 and 28.3 °C) as potentially interacting factors. We used this analysis to determine (i) whether
329 interaction strengths differed between snails that originated from the warm or cold streams

330 (e.g. a main effect of ‘origin’); (ii) whether interaction strengths were temperature dependent
331 (e.g. a main effect of ‘temperature’); and (iii) whether the temperature dependence of
332 interaction strength differed between the snails from the cold and warm streams (e.g.
333 interaction between ‘origin’ and ‘temperature’).

334 We examined the relative contributions of short - and long-term responses to
335 warming: The design of the reciprocal transplant experiment enabled us to disentangle short-
336 term temperature responses attributable to acclimation (e.g. responses to the temperature in
337 the ‘transplanted’ stream) from those reflecting processes operating over longer, time scales
338 (e.g. adaptation to the stream of ‘origin’). Note that these ‘long-term’ effects, which we call
339 ‘adaptation’, could reflect strict genetic microevolution (e.g. resulting in divergent genotypes
340 among populations) or they could represent non-genetic effects of the different temperature
341 regimes that manifest over ontogenetic development, but are nevertheless adaptive
342 (Bonduriansky *et al.*, 2011). In the second GLM we included ‘interaction strength’ as the
343 response variable and ‘timescale’ (‘short’ or ‘long’) and ‘transplant temperature’ (17.5 and
344 28.3 °C) as potentially interacting factors. Here, ‘short-term’ temperature responses were
345 characterised as the change in interaction strength between the stream of origin and the
346 transplant stream. By contrast, the ‘long-term’ temperature response was characterised as the
347 change in interaction strength comparing measurements made only when the snails were in
348 their stream of origin. For better comparison of the steepness of the respiration reaction
349 norms, we re-express the transplant temperature data as Boltzmann temperatures $\left(\frac{1}{kT_c} - \frac{1}{kT}\right)$
350 so that the coefficients of the model yield activation energies in units of eV (see Eq. (1)). In
351 this analysis, a significant interaction between ‘transplant temperature’ and ‘timescale’ would
352 demonstrate that the temperature dependence of interaction strength differs between the
353 ‘short-term’ (E_{short} , change in interaction strength between the stream of origin and the
354 transplant stream, see also Fig. 3), and ‘long-term’ (E_{long} , i.e. change in interaction strength

355 comparing measurements made only when the snails were in their stream of origin, see also
356 Fig. 3). We assume that E_{short} captures rapid physiological plasticity (e.g. acclimation) in
357 interaction strength in response to a change in temperature and E_{long} captures processes
358 operating over longer timescales – e.g. genetic microevolution and non-genetic
359 developmental effects. Consequently, the component of the temperature sensitivity
360 attributable to ‘adaptation’ (recognising that this might be genetically and/or developmentally
361 determined) is given by $E_{\text{adapt}} = E_{\text{long}} E_{\text{short}}$.

362

363 **RESULTS**

364 **Metabolic thermal response curves**

365 The allometric scaling coefficient, α , and the apparent activation energy, E_a , were consistent
366 between the populations of *R. balthica* from the cold and warm streams (see Table 2 for
367 model comparison and estimated parameter values). The temperature normalised rate of
368 respiration, $\ln b(T_c)$, and T_h (the temperature at which respiration was 50% inactivated) were
369 both higher in the population of *R. balthica* from the warm stream. Because the optimum
370 temperature, T_{opt} , depends strongly on T_h (see Eq. (2)), T_{opt} was higher in *R. balthica* from
371 the warmer stream ($T_{\text{opt}} \text{ warm} = 38.25 \pm 0.6 \text{ }^\circ\text{C}$; $T_{\text{opt}} \text{ cold} = 33.05 \pm 1.5 \text{ }^\circ\text{C}$). As $\ln b(T_c)$ and
372 T_{opt} were both higher, the warm populations of *R. balthica* had elevated metabolic rates
373 across the full range of measurement temperatures (Fig. 2).

374

375 **Local adaptation of interaction strength**

376 Interaction strength increased with elevated transplant temperature for the populations of *R.*
377 *balthica* from both the warm and the cold streams (Fig. 3; main effect of ‘transplant
378 temperature’ GLM $t_{1,21} = 2.56$; $P < 0.01$). Furthermore, interaction strengths were
379 consistently higher for the populations of *R. balthica* from the warm stream in both transplant

380 temperatures (Fig. 3; GLM main effect of ‘origin’ $t_{1,121} = 2.92$; $P < 0.005$). These findings are
381 consistent with the higher respiration rates observed in the warm population (Fig. 2) and
382 highlight the association between metabolism and interaction strength.

383

384 **Disentangling the shortand long-term effects of warming on interaction strengths**

385 Our experimental design enabled us to compare temperature sensitivities that capture short-
386 term thermal acclimation (e.g. changes in interaction strength in response to the reciprocal
387 transplant) as well as the long-term temperature sensitivity, which also includes effects of
388 local adaptation (e.g. changes in rates between warm and cold populations quantified in the
389 stream of origin). We found that interaction strength increased with temperature in both the
390 shortand the long-term (Fig. 3). However, the magnitude of the temperature response was
391 significantly larger in the long-term (Fig. 3; interaction between ‘transplant temperature’ and
392 ‘timescale’ on interaction strength; $GLM_{1,18} = -2.91$; $p < 0.05$), where, the average E_{short} was
393 0.46 eV, while E_{long} was significantly higher at 0.99 eV. This divergence between the
394 shortand long-term temperature sensitivities implies a non-trivial contribution of adaptation
395 in amplifying the effects of temperature on interaction strength *in situ*, with the contribution
396 of E_{adapt} of 0.51 and 0.53 eV in the cold and warm adapted populations respectively.

397

398 **DISCUSSION**

399 Understanding how global warming will affect the strength of consumer-resource interactions
400 and the stability of aquatic food webs is a fundamental challenge in evolutionary ecology.

401 Tackling this challenge requires insight on the short-term effects of temperature on
402 metabolism and interaction traits and on how they are modulated by evolutionary and
403 developmental processes over longer time scales. There is evidence from terrestrial (Rall *et al.*,
404 2010; Barton, 2011; Vucic-Pestic *et al.*, 2011; Brose *et al.*, 2012), freshwater (Kratina *et al.*
405 *et al.*, 2012) and marine ecosystems (Sanford, 1999), that warming is likely to increase the

406 strength of consumer-resource interactions, at least in the short-term, owing to the
407 exponential effects of temperature on the consumption rates of mobile ectothermic consumers
408 (Dell *et al.*, 2014; Gilbert *et al.*, 2014). What is less clear however, is how long-term
409 responses to rising temperatures will modulate the direct effects of warming on species
410 interactions. Space-for-time substitutions across broad spatial scales indicate that local
411 adaptation to different thermal regimes can play an important role in shaping species
412 interactions, often compensating for the direct effects of temperature on interaction traits
413 (Barton, 2011; De Block *et al.*, 2012). Here, we build on this work by investigating the
414 effects of temperature and local adaptation on the interaction between the gastropod grazer,
415 *R. balthica*, and its algal resource. Our study contributes novel insights in a number of ways.
416 First, we explore patterns of local adaptation over a relatively small spatial scale (m as
417 opposed to km). The two streams in our experiment are separated by approximately 500 m
418 but differ in temperature by 11 °C. Dispersal, gene flow and genetic divergence among
419 populations in this species are strongly related to geographic distance (Johansson *et al.*,
420 2016), and *R. balthica* are known to disperse slowly, but regularly over short distances
421 (Kappes & Haase 2012, Hoffman *et al.* 2006), but rarely over long distances, where such
422 events are usually associated with passive dispersal, e.g. via association with vertebrates
423 (Bolotov *et al.*, 2017, Hansson & Akesson., 2014). Thus, our study over a relatively small
424 spatial scale (500m), provides insight into how metabolic and resource acquisition traits in
425 closely related natural populations have diverged in response to warming and is therefore
426 directly relevant for understanding the effects of rapid climate change (Richter-Boix *et al.*,
427 2010; Keller *et al.*, 2013; Merilä & Hendry, 2014). Second, we quantified the effects of
428 temperature on both metabolic and consumption rates to determine how temperature-driven
429 selection on key traits shape the effects of long-term warming on the strength of consumer-
430 resource interactions.

431 We found significant variation in the thermal response curves for respiration between
432 the populations of *R. balthica* from the warm and cold streams. The optimum temperature
433 (T_{opt}) for respiration was higher in the warm population (i.e. metabolic rates peaked at higher
434 temperatures). Furthermore, the inactivation energy (E_h) was lower in the warm population,
435 indicating that declines in the rate of respiration after the optimum (i.e. at high temperatures)
436 were less pronounced than in grazers from the cold stream, where metabolic rates peaked at
437 lower temperatures and declined markedly at temperatures above T_{opt} . These divergences
438 indicate that the different thermal regimes in these streams have selected for different
439 metabolic traits in warm and cold populations of *R. balthica*. Whilst the higher T_{opt} and lower
440 E_h in the warm population were in line with expectations assuming local thermal adaptation,
441 we found no evidence that metabolic performance at high temperature was traded-off against
442 performance at low temperature. Instead, metabolic rates were higher for *R. balthica* from the
443 warm stream across all measurement temperatures. These results are in broad agreement with
444 the “hotter is better” hypothesis, which proposes that maximal performance of organisms
445 with higher optimal temperatures should be greater than those with lower optimum
446 temperatures because of the thermodynamic constraints imposed by high temperatures on
447 enzyme kinetics (Knies *et al.*, 2009, Huey & Kingsolver, 1993; Kingsolver *et al.*, 2004;
448 Angilletta *et al.*, 2010). Indeed maximal respiration rates in the population from the warm
449 stream were greater than those from the cool (ln(R) warm stream: $3.39 \pm 0.14 \mu\text{mol O}_2 \text{ L}^{-1} \text{ h}^{-1}$,
450 1 , and cool stream: $2.54 \pm 0.26 \mu\text{mol O}_2 \text{ L}^{-1} \text{ h}^{-1}$, both $\pm 1\text{s.e.m.}$). The lower E_h (i.e. the
451 steepness of the decline of the thermal reaction norm past the optimum), and higher $\ln b(T_c)$,
452 i.e. the rate of respiration normalised to 18 °C, in the warm population also meant that the
453 thermal response curve for *R. balthica* from the warm stream was broader. In agreement with
454 previous work (e.g. on bacteriophages, (Knies *et al.*, 2009, Huey & Kingsolver, 1993;
455 Kingsolver *et al.*, 2004; Angilletta *et al.*, 2010), our data for the gastropod *R. balthica*

456 indicate that adaptation to higher temperatures resulted in both greater maximal metabolic
457 performance and a broader metabolic thermal reaction norm.

458 The general patterns observed in the metabolic traits were also reflected in the effects
459 of temperature on interaction strength. Interaction strength was higher for individuals placed
460 in the warm stream, irrespective of their stream of origin. These findings suggest that
461 elevated temperatures increase consumption rates though the effects of temperature on
462 respiratory physiology, but local adaptation to warmer environments also results in a
463 correlated increase in metabolism and interaction strength at low temperature. This may have
464 important wider implications for the effects of warming on the structure, functioning and
465 stability of aquatic food webs (Rall *et al.*, 2010; O'Connor *et al.*, 2011; Vucic-Pestic *et al.*,
466 2011; Dell *et al.*, 2014; Fussmann *et al.*, 2014; Gilbert *et al.*, 2014). If long-term responses to
467 increasing temperature give rise to higher maximal rates of metabolism and consumption as
468 well as elevating rates at lower temperatures, then the effects of warming on the strength of
469 consumer-resource interactions in the long-term could be greater than previously anticipated
470 (Gilbert *et al.*, 2014). Indeed, work on experimental warming of aquatic ecosystems has
471 shown that increases in the strength of top-down control can have profound effects on
472 community structure and ecosystem processes (Burgmer & Hillebrand, 2011; Kratina *et al.*,
473 2012; Yvon-Durocher *et al.*, 2015). Elevated grazing rates at warmer temperatures can have a
474 wide range of impacts in aquatic systems, with evidence for both increases (Yvon-Durocher
475 *et al.*, 2015) and decreases (Burgmer & Hillebrand, 2011) in algal species richness, biomass
476 and productivity.

477 In our experiments, the thermal sensitivities of metabolic rates were much larger than
478 those of interaction strengths in the short-term (e.g. 0.96 and 0.45 eV respectively), in line
479 with findings in other invertebrate systems (Rall *et al.*, 2010; Vucic-Pestic *et al.*, 2011;
480 Fussmann *et al.*, 2014). These findings suggest that rates of grazing and metabolism were

481 clearly linked, but became decoupled when individuals experience rapid changes in
482 temperature that depart substantially from those in their local environment. In the short-term,
483 if increases in metabolic demands with temperature are greater than those of consumption
484 rates (as found here), then less energy will be transferred from the resource to the consumer,
485 i.e. more is lost through respiration, (see also Rall *et al.* 2010). If such imbalances are
486 maintained over long periods of time then starvation of the consumer can ultimately result in
487 a decline in top-down control on the resource (Rall *et al.* 2010, Fussmann *et al.*, 2014; Binzer
488 *et al.*, 2015). However, when consumers' feeding rates are more sensitive to temperature than
489 metabolic rates, interaction strengths can become amplified in warmer environments, leading
490 to faster resource depletion and eventually driving either the resource or the consumer to
491 local extinction (Vasseur & McCann, 2005). Long-term effects of temperature on interaction
492 strengths have so far only been explored using food web models, parameterised using
493 temperature sensitivities derived from short-term experiments (Vasseur & McCann, 2005;
494 Rall *et al.*, 2012; Fussmann *et al.*, 2014). Consequently, such analyses do not capture the
495 evolutionary and developmental effects which can modulate the short-term effects of
496 temperature on *per capita* rates. Our results highlight substantial differences between the
497 short and long-term effects of temperature on interaction strength; implying that local
498 adaptation can play an important role in modulating the balance between metabolic and
499 consumption rates.

500 We quantified the short and long-term effects of temperature in the reciprocal
501 transplant experiment. The short-term temperature response (E_{short}) captures the effects of
502 physiological plasticity over the 48h experiment. Conversely, the long-term response (E_{long})
503 also accounts for processes operating over longer timescales, including genetic micro
504 evolution and non-genetic developmental effects of temperature. In our experiment, E_{long} was
505 higher than E_{short} , implying a significant role for long-term processes in shaping the effects of

506 temperature on *in situ* interaction strengths. Notably, the higher E_{long} was driven both by
507 elevated grazing rates in the warm populations in the warm stream and lower rates in the cold
508 populations in the cold stream. These results diverge from our expectations based on the
509 metabolic cold adaptation hypothesis (Addo-Bediako *et al.*, 2000) which would predict
510 populations from warmer environments should dampen the acute effects of temperature on
511 metabolic rates. On the contrary, our results suggest that adaptation to warming amplified the
512 effects of temperature on metabolic as well as grazing rates. The lower interaction strengths
513 in the population of *R. balthica* from the colder stream highlight unexpected long-term
514 effects of temperature on species interactions. For example, maintenance of low grazing rates
515 in the cold stream could arise via differences in consumer and/or resource stoichiometry
516 between warmed and cold streams, such that lower consumption rates are required in the cold
517 stream in order to achieve stoichiometric homeostasis (Cross *et al.*, 2005; 2015) under the
518 prevailing temperature regime. Thus, understanding the impacts of environmental change on
519 the strength of consumer-resource interactions over timescales that are relevant to the rate of
520 climate change (e.g. gradual warming over decades) will require an appreciation both of the
521 direct effects of rising temperatures on species interactions and the reciprocal feedback
522 between ecological and evolutionary dynamics (Fussmann *et al.*, 2007; Gravel *et al.*, 2010;
523 Loeuille, 2010; Urban, 2013; Barraclough, 2015)

524

525 We used a natural geothermal temperature gradient to investigate how warming influences
526 the strength of algal-grazer interactions via the direct effects of temperature on metabolism
527 and consumption, and indirect feedbacks through adaptation. Metabolic rates and interaction
528 strength increased with temperature in the same way for both the warm and cold populations
529 of *R. balthica*, suggesting that rapid changes in temperature have a consistent effect on
530 interactions between mobile consumers and sessile resources, mediated by the effects of

531 temperature on consumer metabolic rates (Dell *et al.*, 2014). However, the warm populations
532 had higher metabolic and grazing rates across all measurement temperatures compared to
533 their colder counterparts. These findings are consistent with the ‘hotter is better and broader’
534 hypothesis (Huey & Kingsolver, 1993; Knies *et al.*, 2009; Angilletta *et al.*, 2010) (e.g.
535 adaptation to warming gives rise to higher maximal metabolic rates and broader thermal
536 reaction norms). In consequence, our results suggest that warming could increase the strength
537 of algal-grazer interactions, which are often ‘keystone’ interactions in aquatic systems, both
538 via the thermodynamic effects of higher temperatures on enzyme kinetics and through
539 correlated increases in *per capita* metabolism and consumption as organisms adapt to warmer
540 temperatures.

541

542 **Conflict of interest**

543 The authors declare no conflict of interest

544

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550 **References**

- 551 Addo-Bediako A, Chown SL, Gaston KJ (2000) Thermal tolerance, climatic variability, and
552 latitude. *Proceedings of the Royal Society B: Biological Sciences*, **267**, 739–745.
- 553 Allen, A. P., J. F. Gillooly, and J. H. Brown. 2005. Linking the global carbon cycle to
554 individual metabolism. *Functional Ecology* 19:202–213.
- 555 Angilletta MJ Jr, Huey RB, Frazier MR (2010) Thermodynamic Effects on Organismal
556 Performance: Is Hotter Better? *Physiological and Biochemical Zoology*, **83**, 197–206.
- 557 Angilletta MJ Jr, Wilson RS, Navas CA, James RS (2003) Tradeoffs and the evolution of
558 thermal reaction norms. *Trends in Ecology & Evolution*, **18**, 234–240.
- 559 Barraclough TG (2015) How Do Species Interactions Affect Evolutionary Dynamics Across
560 Whole Communities? *Annual Review of Ecology, Evolution, and Systematics*, **46**, 25–48.
- 561 Barton BT (2011) Local adaptation to temperature conserves top-down control in a grassland

562 food web. *Proceedings of the Royal Society B: Biological Sciences*, **278**, 3102–3107.

563 Berlow EL, Neutel AM, Cohen JE et al. (2004) Interaction strengths in food webs: issues and
564 opportunities. *Journal of Animal Ecology*, **73**, 585–598.

565 Binzer A, Guill C, Rall BC, Brose U (2015) Interactive effects of warming, eutrophication
566 and size structure: impacts on biodiversity and food-web structure. *Global Change
567 Biology*, **22**, 220–227.

568 Bolotov IN, Kondakov AV , Vikhrev IV , Aksenova OV , Bespalaya YV , Gofarov MY ,
569 Kolosova1 YS, Konopleva E , Spitsyn VM, Tanmuangpak K & Tumpeesuwan S (2017)
570 Ancient River Inference Explains Exceptional Freshwater Mussel Radiations, *Scientific
571 Reports*, **7**: 2135 | doi:10.1038/s41598-017-02312-z

572 Bonduriansky R, Crean AJ, Day T (2011) The implications of nongenetic inheritance for
573 evolution in changing environments. *Evolutionary Applications*, **5**, 192–201.

574 Brose U, Dunne JA, Montoya JM, Petchey OL, Schneider FD, Jacob U (2012) Climate
575 change in size-structured ecosystems. *Philosophical Transactions of the Royal Society B:
576 Biological Sciences*, **367**, 2903–2912.

577 Brown JH, Gillooly JF, Allen AP, Van M Savage, West GB (2004) Toward A Metabolic
578 Theory Of Ecology. *Ecology*, **85**, 1771–1789.

579 Burgmer T, Hillebrand H (2011) Temperature mean and variance alter phytoplankton
580 biomass and biodiversity in a long-term microcosm experiment. *Oikos*, **120**, 922–933.

581 Burgmer T, Reiss J, Wickham SA, Hillebrand H (2010) Effects of snail grazers and light on
582 the benthic microbial food web in periphyton communities. *Aquatic Microbial Ecology*,
583 **61**, 163–178.

584 Cross WF, Benstead JP, Frost PC, Thomas SA (2005) Ecological stoichiometry in freshwater
585 benthic systems: recent progress and perspectives. *Freshwater Biology*, **50**, 1895–1912.

586 Cross WF, Hood JM, Benstead JP, Huryn AD, Nelson D (2014) Interactions between
587 temperature and nutrients across levels of ecological organization. *Global Change
588 Biology*, **21**, 1025–1040.

589 De Block M, Pauwels K, Van Den Broeck M, De Meester L, Stoks R (2012) Local genetic
590 adaptation generates latitude-specific effects of warming on predator-prey interactions.
591 *Global Change Biology*, **19**, 689–696.

592 Dell AI, Pawar S, Savage VM (2014) Temperature dependence of trophic interactions are
593 driven by asymmetry of species responses and foraging strategy. *J Anim Ecol*, **83**, 70–84.

594 Dell AI, Pawar S, Savage VM (2011) Systematic variation in the temperature dependence of
595 physiological and ecological traits. *Proceedings of the National Academy of Sciences*,
596 **108**, 10591–10596.

597 Deutsch CA, Tewksbury JJ, Huey RB, Sheldon KS, Ghalambor CK, Haak DC, Martin PR
598 (2008) Impacts of climate warming on terrestrial ectotherms across latitude. *Proceedings
599 of the National Academy of Sciences*, **105**, 6668–6672.

600 Englund G, Öhlund G, Hein CL, Diehl S (2011) Temperature dependence of the functional
601 response. *Ecology Letters*, **14**, 914–921.

602 Eyring, H. (1935). The activated complex in chemical reactions. *The Jour- nal of Chemical
603 Physics*, **3**, 107–115.

604

605 Fukami T, Wardle DA (2005) Long-term ecological dynamics: reciprocal insights from
606 natural and anthropogenic gradients. *Proceedings of the Royal Society B: Biological
607 Sciences*, **272**, 2105–2115.

608 Fussmann GF, Loreau M, Abrams PA (2007) Eco-evolutionary dynamics of communities
609 and ecosystems. *Functional Ecology*, **21**, 465–477.

610 Fussmann KE, Schwarzmüller F, Brose U, Jousset A, Rall BC (2014) Ecological stability in
611 response to warming. *Nature Climate Change*, **4**, 206–210.

- 612 Gilbert B, Tunney TD, McCann KS et al. (2014) A bioenergetic framework for the
613 temperature dependence of trophic interactions (ed Wootton T). *Ecology Letters*, **17**,
614 902–914.
- 615 Gillooly JF (2001) Effects of Size and Temperature on Metabolic Rate. *Science*, **293**, 2248–
616 2251.
- 617 Gravel D, Bell T, Barbera C, Bouvier T, Pommier T, Venail P, Mouquet N (2010)
618 Experimental niche evolution alters the strength of the diversity-productivity
619 relationship. *Nature*, **469**, 89–92.
- 620 Gudmundsdottir, R., S. Pálsson, E. R. Hannesdottir, J. S. Olafsson, G. M. Gislason, and B.
621 Moss. 2013. Diatoms as indicators: The influences of experimental nitrogen enrichment
622 on diatom assemblages in sub-Arctic streams. *Ecological Indicators* 32:74–81.
- 623 Hansson LA&Akesson S Editors (2014) *Animal Movement Across Scales*, Oxford University
624 Press
- 625 Hoffman AL, Olden JD, Monroe JB, LeRoy Poff N, Wellnitz T, Wiens JA (2006) Current
626 velocity and habitat patchiness shape stream herbivore movement. *Oikos* **115**, 358–368
- 627 Hoffmann AA, Sgrò CM (2011) Climate change and evolutionary adaptation. *Nature*, **470**,
628 479–485.
- 629 Holm-Hansen O, Riemann B (1978) Chlorophyll a Determination: Improvements in
630 Methodology. *Oikos*, **30**, 438.
- 631 Huey RB, Kingsolver JG (1993) Evolution of Resistance to High Temperature in Ectotherms.
632 *The American Naturalist*, **142**, S21–S46.
- 633 Johansson MP, Quintela M, Laurila A (2016) Genetic divergence and isolation by thermal
634 environment in geothermal populations of an aquatic invertebrate. *Journal of*
635 *Evolutionary Biology*, **29(9)**:1701-12
- 636 Kappes H, Haase, P, (2012) Slow, but steady: dispersal of freshwater molluscs, *Aquatic*
637 *Sciences*, **74 (1)**, 1-14
- 638 Keller I, Alexander JM, Holderegger R, Edwards PJ (2013) Widespread phenotypic and
639 genetic divergence along altitudinal gradients in animals. *Journal of Evolutionary*
640 *Biology*, **26**, 2527–2543.
- 641 Kilham SS, Kreeger DA, Lynn SG, Goulden CE, Herrera L (1998) COMBO: a defined
642 freshwater culture medium for algae and zooplankton. *Hydrobiologia*, **377**, 147–159.
- 643 Kingsolver JG, Huey RB (2008) Size, temperature, and fitness: three rules. *Evolutionary*
644 *Ecology Research*, **10**, 251–268.
- 645 Kingsolver JG, Ragland GJ, Shlichta JG (2004) Quantitative genetics of continuous reaction
646 norms: thermal sensitivity of caterpillar growth rates. *Evolution*.
- 647 Knies JL, Kingsolver JG, Burch CL (2009) Hotter Is Better and Broader: Thermal Sensitivity
648 of Fitness in a Population of Bacteriophages. *The American Naturalist*, **173**, 419–430.
- 649 Kratina P, Greig HS, Thompson PL (2012) Warming modifies trophic cascades and
650 eutrophication in experimental freshwater communities. *Ecology*, **93**, 1421–1430.
- 651 Lang B, Rall BC, Brose U (2011) Warming effects on consumption and intraspecific
652 interference competition depend on predator metabolism. *Journal of Animal Ecology*, **81**,
653 516–523.
- 654 Laufkötter C, Vogt M, Gruber N et al. (2015) Drivers and uncertainties of future global
655 marine primary production in marine ecosystem models. *Biogeosciences*, **12**, 6955–6984.
- 656 Loarie SR, Duffy PB, Hamilton H, Asner GP, Field CB, Ackerly DD (2009) The velocity of
657 climate change. *Nature*, **462**, 1052–1055.
- 658 Loeuille N (2010) Influence of evolution on the stability of ecological communities. *Ecology*
659 *Letters*, **13**, 1536–1545.
- 660 May RM (1973) Qualitative stability in model ecosystems. *Ecology*, **54**, 638–641.
- 661 McCann K, Hastings A, Huxel GR (1998) Weak trophic interactions and the balance of

662 nature. *Nature*, **395**, 794–798.

663 Merilä J, Hendry AP (2014) Climate change, adaptation, and phenotypic plasticity: the
664 problem and the evidence. *Evolutionary Applications*, **7**, 1–14.

665 O'Connor MI (2009) Warming strengthens an herbivore–plant interaction. *Ecology*, **90**, 388–
666 398.

667 O'Connor MI, Gilbert B, Brown CJ (2011) Theoretical predictions for how temperature
668 affects the dynamics of interacting herbivores and plants. *The American Naturalist*, **178**,
669 626–638.

670 OGorman EJ, Pichler DE, Adams G et al. (2012) *Impacts of Warming on the Structure and*
671 *Functioning of Aquatic Communities: Individual to Ecosystem-Level Responses*, 1st edn,
672 Vol. 47. Elsevier Ltd., 96 p.

673 Otto SB, Rall BC, BROSE U (2007) Allometric degree distributions facilitate food-web
674 stability. *Nature*, **450**, 1226–1229.

675 Padfield D, Lowe C, Buckling A, French-Constant R, Jennings A, Shelley F, Ólafsson JS,
676 Yvon-Durocher G (2017) Metabolic compensation constrains the temperature
677 dependence of gross primary production, *Ecology Letters*, **20**, 1250–1260

678 Paine RT (1980) Food webs: linkage, interaction strength and community infrastructure.
679 *Journal of Animal Ecology*, **49**, 666.

680 Vucic-Pestic O, Ehnes RB, Rall BC (2011) Warming up the system: higher predator feeding
681 rates but lower energetic efficiencies. *Global Change Biology*, **17**, 1301–1310

682 Rall BC, Vucic-Pestic O, Ehnes RB, Emmerson M, Brose U (2010) Temperature, predator-
683 prey interaction strength and population stability. *Global Change Biology*, **16**, 2145–
684 2157.

685 Rall BC, Brose U, Hartvig M, Kalinkat G, Schwarzmuller F, Vucic-Pestic O, Petchey OL
686 (2012) Universal temperature and body-mass scaling of feeding rates. *Philosophical*
687 *Transactions of the Royal Society B: Biological Sciences*, **367**, 2923–2934.

688 Ricciardi A, Bourget E (1998) Weight-to-weight conversion factors for marine benthic
689 macroinvertebrates. *Marine Ecology Progress Series*, **171**, 245–251.

690 Richter-Boix A, Teplitsky C, Rogell B, Laurila A (2010) Local selection modifies phenotypic
691 divergence among *Rana temporaria* populations in the presence of gene flow. *Molecular*
692 *Ecology*, **19**, 716–731.

693 Sanford E (1999) Regulation of keystone predation by small changes in ocean temperature.
694 *Science*, **283**, 2095–2097.

695 Schoolfield RM, Sharpe PJH, Magnuson CE (1981) Non-linear regression of biological
696 temperature-dependent rate models based on absolute reaction-rate theory. *Journal of*
697 *Theoretical Biology*, **88**, 719–731.

698 Sentis A, Morisson J, Boukal DS (2015) Thermal acclimation modulates the impacts of
699 temperature and enrichment on trophic interaction strengths and population dynamics.
700 *Global Change Biology*, **21**, 3290–3298.

701 Sharpe PJ, DeMichele DW (1977) Reaction kinetics of poikilotherm development. *Journal of*
702 *Theoretical Biology*, **64**, 649–670.

703 Shurin JB, Clasen JL, Greig HS, Kratina P, Thompson PL (2012) Warming shifts top-down
704 and bottom-up control of pond food web structure and function. *Philosophical*
705 *Transactions of the Royal Society B: Biological Sciences*, **367**, 3008–3017.

706 Skalski GT, Gilliam JF (2001) Functional Responses With Predator Interference: Viable
707 Alternatives To The Holling Type II Model. *Ecology*, **82**, 3083.

708 Urban MC (2013) Evolution mediates the effects of apex predation on aquatic food webs.
709 *Proceedings of the Royal Society B: Biological Sciences*, **280**, 20130859–20130869.

710 Van M Savage, Gillooly JF, Brown JH, West GB, Charnov EL (2015) Effects of Body Size
711 and Temperature on Population Growth. *The American Naturalist*, **163**, 429–441.

- 712 Vasseur DA, McCann KS (2005) A mechanistic approach for modeling temperature-
713 dependent consumer-resource dynamics. *The American Naturalist*, **166**, 184–198.
- 714 West DC, Post DM (2016) Impacts of warming revealed by linking resource growth rates
715 with consumer functional responses (ed Behmer S). *Journal of Animal Ecology*, **85**, 671–
716 680.
- 717 West-Eberhard MJ. (2003). *Developmental plasticity and evolution*. Oxford University Press.
- 718 Yvon-Durocher G, Allen AP, Cellamare M et al. (2015) Five Years of Experimental
719 Warming Increases the Biodiversity and Productivity of Phytoplankton (ed Levin SA).
720 *PLoS Biology*, **13**, e1002324–22.
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723 **Tables and Table captions:**

724

725 **Table 1. Physical and chemical characteristics of the streams.** Temperature data were
 726 collected over a 3-day period. All other parameters were collected on the first day of the day
 727 of the experiment. Temperature data are displayed as means \pm 1SD. All other data were
 728 originally collected for correlation with temperature across the catchment area (all 11
 729 streams), so that replication for these parameters was on the level of stream identity.
 730

Parameter	Stream 5	Stream 11
Average temperature ($^{\circ}\text{C}$, 5 days)	17.5 \pm 4.5	28.3 \pm 1.3
pH	7.63	7.17
Conductivity (μS)	273.6	235.7
NO_2 ($\mu\text{mol L}^{-1}$)	0.22	0.24
NO_3 ($\mu\text{mol L}^{-1}$)	0.57	0.29
NH_4 ($\mu\text{mol L}^{-1}$)	0.17	0.19
PO_4 ($\mu\text{mol L}^{-1}$)	0.27	0.35
Velocity (m s^{-1})	0.10	0.15
Stream depth (m)	0.041	0.042

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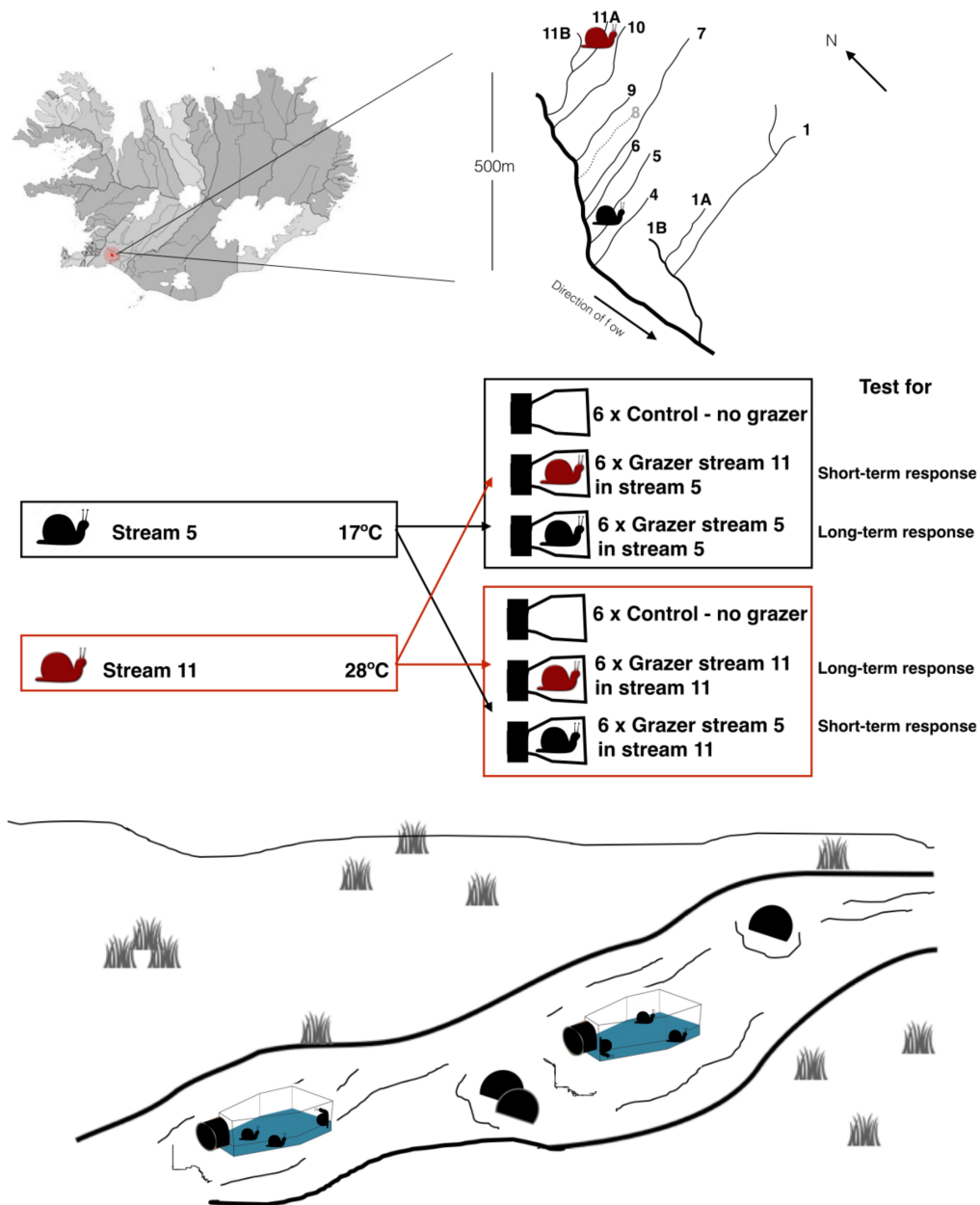
Table 2. Parameter estimates and output from the best fitting gnls model to the thermal response curves of respiration rates. Differences in treatments are given in **bold**. Parameter estimates are taken from the averaged generalised linear models along with their standard errors (± 1 s.e.m). C = cold stream. W = warm stream. See Supporting Information for details on model selection and information on AICc scores for all possible models. Here, the model average of the conditional average output for the four best models (within 2 AICc units of each other) is displayed.

Non-linear mixed model output for respiration rates (R)		
Treatment effect on	Estimate	± 1 s.e.m.
E_a ln $R(T_c)$	C: 0.96	0.05
	C: 1.77	0.15
	W: 2.04	0.12
E_h	C: 5.01	0.97
	W: 3.16	0.96
T_h [K] (°C)	C: 307.16 (34.01)	0.94
	W: 314.15	1.69
	(41.00)	0.78
α	0.36	0.03

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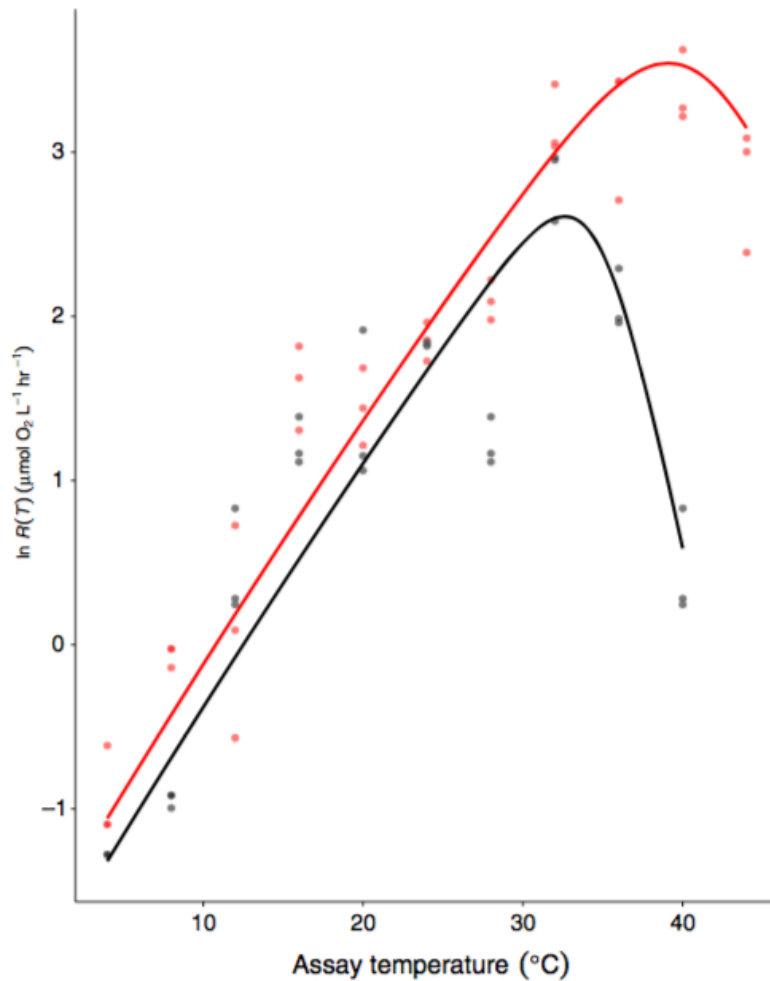
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Figure Captions online



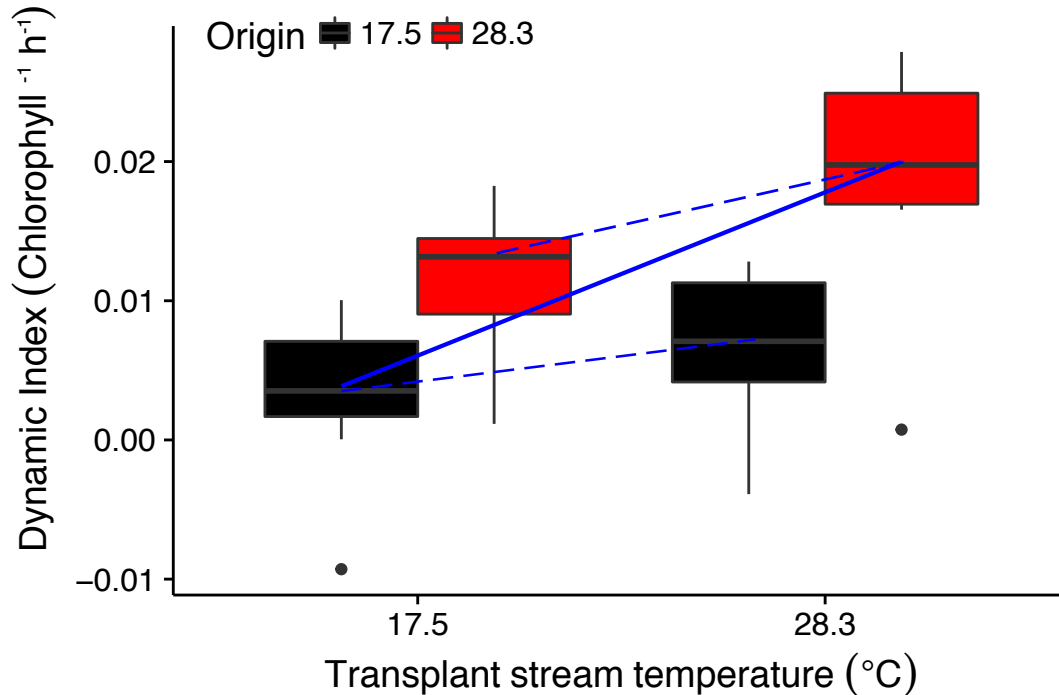
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FIGURE 1 Map and experimental set-up. Top panel: The catchment area, with streams used in this experiment indicated by black (for the colder stream 5 with $17.5\text{ }^{\circ}\text{C} \pm 4.5\text{ }^{\circ}\text{C}$) and red (for the warmer stream 11A with $28.3\text{ }^{\circ}\text{C} \pm 1.3\text{ }^{\circ}\text{C}$) snail icons. Middle panel: Schematic overview of experimental set-up for the grazing experiment. Lower panel: Schematic overview of microcosms in stream. Tent pegs and cable ties used for fixing the microcosms in place have been omitted from the schematic. See Figure S1 for a photograph.



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FIGURE 2 Thermal response curves for respiration. Thermal response curves of respiration rates in $\mu\text{mol O}_2 \text{ L}^{-1} \text{ h}^{-1}$ as a function of increasing temperature for populations of the snail *Radix balthica* from the cold (black) and warm (red) stream. Lines are derived from fitting a modified Sharpe-Schoolfield equation (see methods) to the rate data. Snails from the warm stream have higher temperature normalised metabolic rates ($\ln R(T_c)$, with $T_c=18^\circ\text{C}$) at all measurement temperatures and have higher optimal temperatures (T_{opt}), than snails from the cold stream. The inactivation energy (E_h) is lower in snails from the warm stream, resulting in a curve that is both broader and elevated in comparison to the thermal response curve of respiration for snails from the cold stream.



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FIGURE 3 Long-term and short-term effects of stream temperature on interaction strength Long-term and short-term effects of temperature in interaction strength measured via the dynamic index in units of chlorophyll consumed per hour. Populations originating from the warm stream have stronger interaction strength indices in all environments and the highest dynamic index overall was found for snails from the warm stream in their original environment. Interaction strength increased with temperature both in the short-term (E_{short} , dashed blue lines) and in the long-term (E_{long} , solid blue line, comparing dynamic indexes of snails in their temperatures of origin), with E_{long} significantly greater than E_{short} .

799 **Figure Captions print (greyscale)**

800 **FIGURE 1 Map and experimental set-up. Top panel:** The catchment area, with streams used in this experiment indicated by black (for the colder stream 5 with $17.5\text{ }^{\circ}\text{C} \pm 4.5\text{ }^{\circ}\text{C}$) and grey (for the warmer stream 11A with $28.3\text{ }^{\circ}\text{C} \pm 1.3\text{ }^{\circ}\text{C}$) snail icons. **Middle panel:** Schematic overview of experimental set-up for the grazing experiment. **Lower panel:** Schematic overview of microcosms in stream. Tent pegs and cable ties used for fixing the microcosms in place have been omitted from the schematic. See Figure S1 for a photograph.

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808 **FIGURE 2 Thermal response curves for respiration.** Thermal response curves of respiration rates in $\mu\text{mol O}_2\text{ L}^{-1}\text{ h}^{-1}$ as a function of increasing temperature for populations of the snail *Radix balthica* from the cold (black, solid line, circles) and warm (grey, dotted line, triangles) stream. Lines are derived from fitting a modified Sharpe-Schoolfield equation (see methods) to the rate data. Snails from the warm stream have higher temperature normalised metabolic rates ($\ln R(T_c)$, with $T_c=18^{\circ}\text{C}$) at all measurement temperatures and have higher optimal temperatures (T_{opt}), than snails from the cold stream. The inactivation energy (E_h) is lower in snails from the warm stream, resulting in a curve that is both broader and elevated in comparison to the thermal response curve of respiration for snails from the cold stream.

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818 **FIGURE 3 Long-term and short-term effects of stream temperature on interaction**
819 **strength** Long-term and short-term effects of temperature in interaction strength measured
820 via the dynamic index in units of chlorophyll consumed per hour. Populations originating
821 from the warm stream (grey) have stronger interaction strength indices than snails from the
822 colder stream (black) in all environments and the highest dynamic index overall was found
823 for snails from the warm stream in their original environment. Interaction strength increased
824 with temperature both in the short-term (E_{short} , dashed lines) and in the long-term (E_{long} , solid
825 line, comparing dynamic indexes of snails in their temperatures of origin), with E_{long}
826 significantly greater than E_{short} .
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