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1 **Coral adaptation to climate change: meta-analysis reveals high heritability across**
2 **multiple traits**

3 **Running Title: Trait heritability in reef-building corals**

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22

23 **Abstract**

24 Anthropogenic climate change is a rapidly intensifying selection pressure on
25 biodiversity across the globe and, particularly, on the world's coral reefs. The rate of
26 adaptation to climate change is proportional to the amount of phenotypic variation that can be
27 inherited by subsequent generations (i.e., narrow-sense heritability, h^2). Thus, traits that have
28 higher heritability (e.g., $h^2 > 0.5$) are likely to adapt to future conditions faster than traits with
29 lower heritability (e.g., $h^2 < 0.1$). Here, we synthesize 95 heritability estimates across 19
30 species of reef-building corals. Our meta-analysis reveals low heritability ($h^2 < 0.25$) of gene
31 expression metrics, intermediate heritability ($h^2 = 0.25\text{--}0.50$) of photochemistry, growth, and
32 bleaching, and high heritability ($h^2 > 0.50$) for metrics related to survival and immune
33 responses. Some of these values are higher than typically observed in other taxa, such as
34 survival and growth, while others were more comparable, such as gene expression and
35 photochemistry. There was no detectable effect of temperature on heritability, but narrow-
36 sense heritability estimates were generally lower than broad-sense estimates, indicative of
37 significant non-additive genetic variation across traits. Trait heritability also varied depending
38 on coral life stage, with bleaching and growth in juveniles generally having lower heritability
39 compared to bleaching and growth in larvae and adults. These differences may be the result
40 of previous stabilizing selection on juveniles or may be due to constrained evolution resulting
41 from genetic trade-offs or genetic correlations between growth and thermotolerance. While
42 we find no evidence that heritability decreases under temperature stress, explicit tests of the
43 heritability of thermal tolerance itself – such as coral thermal reaction norm shape – are
44 lacking. Nevertheless, our findings overall reveal high trait heritability for the majority of
45 coral traits, suggesting corals may have a greater potential to adapt to climate change than has
46 been assumed in recent evolutionary models.

47

48 **Keywords:** adaptation, animal model, Breeder's equation, climate change, evolution, natural
49 selection, Scleractinia, thermal tolerance

50 **Introduction**

51 Anthropogenic climate change is one of the greatest selective pressures on organisms
52 worldwide (Davis et al. 2005; Hughes et al. 2018b; Nolan et al. 2018). To avoid extinction,
53 species need to either acclimatize, move to new habitats, or adapt to new conditions (Davis et
54 al. 2005; Jump and Peñuelas 2005; Gienapp et al. 2008). Acclimatization on its own may
55 initially increase the duration of time that a population can persist in an altered environment,
56 but is bounded by physiological thresholds that limit tolerance of long-term environmental
57 change, and for populations living close to their extremes already (Comte & Olden, 2017;
58 Sasaki & Dam, 2019; van Heerwaarden, Kellermann, & Sgr, 2016). Migration to new
59 environments is similarly limited (Jump & Peñuelas, 2005; Schloss, Nuñez, & Lawler, 2012;
60 Walters & Berger, 2019), especially for predominantly sessile organisms for which range
61 extension depends upon long-distance dispersal of offspring (Hughes et al. 2003;
62 Archambault et al. 2018; c.f. Kremer et al. 2012). Consequently, understanding whether and
63 how species are likely to adapt to future conditions is crucial in predicting species persistence
64 in the context of climate change (C. A. Logan, Dunne, Eakin, & Donner, 2014).

65 Adaptive evolution to a changing environment occurs when population genotype
66 frequencies change to express traits or phenotypes that provide increased fitness (Falconer &
67 Mackay, 1996). However, adaptation of a trait can only occur at a rate proportional to the
68 narrow-sense heritability coefficient, h^2 , calculated as the ratio between population variance
69 attributable to additive genetic effects, V_a , and the total observed phenotypic variance, V_p
70 (Box 1). The narrow-sense heritability coefficient is a key parameter in the univariate
71 ‘Breeder’s equation’, which predicts the mean population response in trait values for a single
72 trait undergoing selection.

73

74

Box 1. Heritability and the Breeder's equation

Heritability is the proportion of a trait's relative genetic variation compared to its total phenotypic variation (V_P). As a dimensionless quantity that describes population responses to selection, heritability is often calculated to compare across different traits, populations, or species (Visscher, Hill, & Wray, 2008). There are two types of heritability: broad-sense heritability (H^2) and narrow-sense heritability (h^2).

$$\text{Broad-sense: } H^2 = \frac{V_A + V_D + V_I}{V_P} \qquad \text{Narrow-sense: } h^2 = \frac{V_A}{V_P} \qquad (1)$$

The former is 'broad' in that it includes all sources of genetic variation (V_G), including additive genetic variation (V_A), as well as non-additive sources such as dominance (V_D) and epistasis (V_I), which are genetic effects not (necessarily) inherited by offspring produced through sexual reproduction (Falconer & Mackay, 1996), and thus broad-sense heritability is calculated using clones or genets. Narrow-sense heritability, h^2 , is the proportion of phenotypic variance that is due to additive genetic variance (V_A) alone, and thus is the strictly 'heritable' genetic component of the trait of interest in sexually-reproducing individuals. Narrow-sense heritability is thus calculated using pedigree information from genetically distinct individuals.

Important caveats when estimating either narrow or broad-sense heritability include controlling for common environment across individuals as well as considering potential genotype-by-environment interactions, parental effects that may alter offspring phenotype, and the potential for epigenetic inheritance (Falconer & Mackay, 1996; Visscher et al., 2008).

Narrow sense heritability can be used to predict and understand population responses to selection. Consider one population where the mean critical thermal maximum (CT_{max}) of the population is 30°C and CT_{max} is highly heritable, e.g. $h^2 =$

0.5. If a temperature anomaly occurs, resulting in the death of the more heat-susceptible individuals and shifting the mean population CT_{max} to 32°C, the univariate Breeder's equation predicts that the mean change in population response (R) in CT_{max} will be:

$$R = h^2 \cdot S = 0.5 \cdot (32^\circ\text{C} - 30^\circ\text{C}) = +1^\circ\text{C} \quad (2)$$

In other words, an increase in the mean selected population CT_{max} of +2°C (S , the 'selection differential') translates to an expected increase in the next generation's average CT_{max} of +1°C (Falconer & Mackay, 1996; Lande, 1979; Lush, 1937). Now consider a second population experiencing the same selection event, but with a relatively low heritability in CT_{max} , e.g. $h^2 = 0.1$. The predicted change in CT_{max} in the next generation would be: $R = 0.1 \cdot (32^\circ\text{C} - 30^\circ\text{C}) = +0.2^\circ\text{C}$, or a five-fold lower response to the same selection pressure. Thus, the relative response of a trait in a population under selection is proportional to the value of the narrow-sense heritability coefficient for said trait, h^2 .

75

76 Different traits often have different heritability coefficients, and may also covary with
77 one another (Wright et al. 2019). Counterintuitively, traits which are tied closely to biological
78 fitness (e.g., life history traits, longevity/survival, fecundity) often have relatively low
79 heritability compared to physiological and behavioural traits, and compared to morphological
80 traits that often have higher heritability (Martins, Kruuk, Llewelyn, Moritz, & Phillips, 2019;
81 Mousseau & Roff, 1987; Price & Schluter, 1991; Wheelwright, Keller, & Postma, 2014). For
82 example, when populations have previously undergone strong stabilizing selection for a trait
83 tied closely to fitness, the narrower range and variance of trait values observed in the
84 population translates to a reduction in the relative contribution of additive genetic effects to
85 total phenotypic variation, and a decrease in the heritability coefficient (Charmantier &

86 Garant, 2005; Teplitsky, Mills, Yarrall, & Merilä, 2009; Wheelwright et al., 2014).
87 Understanding the potential rate and limits to adaptive evolution will therefore require an
88 understanding of heritability across different traits (Wheelwright et al., 2014).

89 Selective pressures differ not only in terms of the trait being examined, but also across
90 life stages/ages, growth forms, and environments (e.g., genotype-by-environment
91 interactions). Thus, heritability should vary across these factors as well (Charmantier &
92 Garant, 2005; Wheelwright et al., 2014; Wilson, Charmantier, & Hadfield, 2008). Early life
93 stages/ages can experience strong stabilizing selection for traits associated with early life
94 fitness, and thus exhibit reduced h^2 for these traits. Increasing importance of environmental
95 effects and acclimation to local environments can also reduce the relative importance of
96 additive genetic variation and thus h^2 at intermediate stages/ages (Charmantier, Perrins,
97 McCleery, & Sheldon, 2006a). Finally, late-acting mutations can accumulate in older
98 individuals to cause age-dependent increases in V_A , and thus h^2 , for traits tied closely with
99 fitness (Charmantier et al., 2006a; Charmantier, Perrins, McCleery, & Sheldon, 2006b;
100 Wilson et al., 2008). Similar selective pressures can result in similar h^2 values for traits of
101 species occupying similar ecological niches. For example, in reef-building corals, colony
102 growth form directly influences individual growth rate, fecundity, and survival (Madin et al.,
103 2020; Pratchett et al., 2015). Tabular coral species (which form large horizontal plates
104 supported by a central stalk) exhibit increased adult mortality relative to other coral growth
105 forms in the same habitat due to their increased mechanical vulnerability (Madin, Baird,
106 Dornelas, & Connolly, 2014). However, it remains unknown whether and how heritability of
107 traits varies among coral species with different growth forms.

108 In the context of climate change, decreasing environmental suitability and increasing
109 selective pressure on traits tied closely to fitness can reduce trait heritability, resulting in a
110 counter-intuitive reduction in the capacity for populations to evolve to environmental change

111 (Charmantier & Garant, 2005; Wheelwright et al., 2014; Wilson et al., 2006). Conversely,
112 other studies have found no distinguishable relationship between h^2 and environmental
113 favourability (Rowinski & Rogell, 2017), and others still identify positive correlations of h^2
114 with increasingly harsh environmental temperatures (Gunay, Alten, & Ozsoy, 2011). Clearly,
115 further research is required to quantify how heritability may change across life stages and
116 environments of the future, especially when attempting to project population outcomes in
117 response to future conditions.

118 Reef-building scleractinian corals are particularly sensitive to climate stressors, as
119 evidenced by coral bleaching during thermal anomalies. Reef diversity and coral cover have
120 declined throughout the 21st century (T. P. Hughes, Kerry, et al., 2018; Pratchett, Hoey,
121 Wilson, Messmer, & Graham, 2011; Wulff, 2006), with 75% of global reefs now being
122 considered threatened (Burke, Reytar, Spalding, & Perry, 2011). Widespread bleaching of
123 coral communities now occurs at temperatures approximately 0.5°C higher than a decade
124 ago, suggesting strong selection for increased thermotolerance worldwide (Guest et al., 2012;
125 Maynard, Anthony, Marshall, & Masiri, 2008; Sully, Burkepile, Donovan, Hodgson, & van
126 Woesik, 2019). However, given the rapid warming of sea surface temperatures and the
127 increase in the frequency and severity of mass bleaching events on coral reefs worldwide (T.
128 P. Hughes, Anderson, et al., 2018; Lough, Anderson, & Hughes, 2018), it remains unclear
129 whether corals can adapt to the prolonged thermal stress they now experience with increasing
130 regularity (Hoegh-Guldberg, Poloczanska, Skirving, & Dove, 2017; Pandolfi, Connolly,
131 Marshall, & Cohen, 2011). Models estimating long-term coral adaptation to climate change
132 have assumed low to medium heritability of thermotolerance (e.g., $h^2 = 0.01-0.50$; Cropp and
133 Norbury 2020; Matz et al. 2020; Logan et al. 2021), despite some evidence of model
134 outcomes being sensitive to the rate of adaptation (Bay, Rose, Logan, & Palumbi, 2017;
135 Cropp & Norbury, 2020), evidence of high heritability for coral survivorship in high

136 temperatures (e.g., $h^2 = 0.75$: Kirk et al. 2018), and the potential for rapid symbiont evolution
137 in response to thermal change (e.g., Chakravarti et al. 2017; Buerger et al. 2020). Therefore,
138 our ability to project the future of coral populations in the context of climate change is
139 critically dependent upon the estimates of trait heritability used in eco-evolutionary models
140 (C. A. Logan et al., 2014; M. L. Logan, Cox, & Calsbeek, 2014; Visser, 2008).

141 Worldwide, reef-building corals are undergoing increasingly strong selection for
142 temperature tolerance due to anthropogenic climate change. Here, we undertake a
143 quantitative meta-analysis of published heritability estimates for reef-building corals to better
144 understand which traits are likely to change most rapidly given increased environmental
145 change. We examine the relative heritability of different trait types such as coral gene
146 expression, bleaching, growth, symbiont community structure, and survival, and investigate
147 potential interactions among life stages and, for experiments that manipulate temperature, to
148 the magnitude of temperature stress to which corals are exposed. We also estimate the
149 relative heterogeneity of heritability estimates, examine differences between narrow-sense vs.
150 broad-sense heritability estimates, and identify how differences in coral growth forms
151 influence estimates of h^2 .

152

153 **Literature Analysis**

154 *Literature search*

155 We undertook an exhaustive literature review to find all possible heritability point
156 estimates and associated measures of sampling variance (e.g., standard errors, confidence or
157 credibility intervals) for scleractinian corals by canvassing two major research databases:
158 Google Scholar and Web of Science. Keyword searches were conducted in October 2020 to
159 identify all studies reporting heritability estimates for corals, and/or their associated
160 symbionts. We searched for studies using the keywords and Booleans: “heritability” AND

161 “coral” OR “familial effects” AND “coral”. We found a total of 16 studies reporting
162 heritability and one study with heritability estimates that could be extracted from the
163 published data. We then mined the references cited within each paper from the initial search
164 to identify two additional studies reporting heritability estimates. We contacted some study
165 authors when text alluded to heritability estimates, but values were not reported in the final
166 manuscript, all of whom kindly contributed those heritability estimates to our analysis.
167 Where possible, we selected single heritability estimates calculated using a fixed effect of
168 temperature or other treatment, rather than taking multiple heritability estimates calculated by
169 splitting the data by each treatment. Many of the studies report multiple heritability estimates
170 for the same or similar traits. For five studies (Kirk, Howells, Abrego, Burt, & Meyer, 2018;
171 Lohr & Patterson, 2017; Manzello et al., 2019; Wright et al., 2019; Zhang, Million, Ruggeri,
172 & Kenkel, 2019), we selected one representative heritability estimate when there were
173 multiple and highly related h^2 estimates (e.g., Kaplan-Meier survival and percent survival;
174 symbiont abundance and chlorophyll A content; total linear extension and net buoyant
175 weight). Finally, we extracted both broad-sense (H^2) and narrow-sense (h^2) heritability
176 estimates for the same trait where both were reported together in order to examine differences
177 between H^2 vs. h^2 ; however, this occurred only for a single study (Carlson, Budd, Lippé, &
178 Andrew, 2011).

179 The above resulted in a total of 103 unique heritability values estimated using a
180 number of methods (see Supplementary Text S1 for a description of the different methods
181 used). Of the 103 total estimates, eight were further excluded on the basis of statistical issues
182 for one of two reasons: (1) there was insufficient variation in relatedness among individuals
183 to properly assess heritability (one study with one estimate), or (2) there was insufficient or
184 expected null variation in the phenotype being examined (two studies, with one and six
185 estimates each). The latter was the case when studies estimated heritabilities associated with

186 coral mortality or bleaching while in ambient conditions, which results in little to no
187 phenotypic variation from which to calculate narrow-sense heritability (i.e., none of the
188 corals bleached or died). This left 95 unique and valid heritability estimates from 19 studies.

189

190 *Pre-processing*

191 Heritability is calculated as a proportion of total phenotypic variation, and thus is
192 constrained to fall between zero and one (Falconer & Mackay, 1996). Because most classical
193 meta-analytical statistical models assume normally-distributed uncertainty, transformation of
194 our estimates prior to meta-analysis was necessary (Lin & Xu, 2020; Wolfgang Viechtbauer,
195 2010). Thus, we converted point estimates of heritabilities and associated standard errors
196 (SE) to 95% confidence intervals, then transformed both the point estimates as well as the
197 upper and lower 95% confidence (or Bayesian credible) limits to the natural logarithmic scale
198 using the transformation:

$$h_T^2 = \ln[h^2 + 0.2] \quad (1)$$

199 with a horizontal displacement of + 0.2 to avoid excluding lower h^2 CIs that had slightly
200 negative values when the point estimate was close to zero (see Supplementary Text S2 and
201 Supplementary Code Documentation A for details).

202 There were four estimates from two studies that did not report any associated SE or
203 CI values, and another three estimates whose lower CI values (when calculated from the SE)
204 were less than -0.2, which prevented their transformation to the \ln -scale. To include these
205 data in the meta-analysis but down-weight their leverage on the overall analysis, we fit a
206 quantile regression through the 95th quantile of transformed SE vs. transformed h^2 (Koenker,
207 2020; Koenker & Hallock, 2001) using only heritability estimates that were able to be
208 transformed to the \ln -scale. value of the heritability on the transformed scale (h^2_T) to predict
209 SE_T values. We then used this fitted equation ($SE_T = 0.255 - 0.452 \cdot h^2_T$) to interpolate missing

210 SE_T values, making the conservative assumption that they would have values at the upper
211 95th quantile (i.e., among the most uncertain estimates).

212

213 *Factors of interest*

214 We identified five explanatory factors present in most studies: trait type, heritability
215 type, life stage, growth form, and temperature manipulation (Table 1). Differences in
216 heritability estimates among specific coral/symbiont species were also of interest; however,
217 most studies examined only a single species and there was little overlap in species across
218 studies, with the exception of a number of studies examining *Acropora millepora* heritability.

219

Table 1. Explanatory factors and covariates examined in the meta-analysis of coral heritability estimates

Factor/Covariate	Levels	Definition
<i>Heritability type</i>	Broad-sense heritability, H^2	The proportion of phenotypic variation explained by all genetic effects, which includes sources of variance associated with additive, dominance, and epistatic effects
	Narrow-sense heritability, h^2	the proportion of phenotypic variation explained by additive genetic effects
<i>Trait type</i>	Gene expression	Up- or down-regulation of various genes involved in intracellular stress pathways
	Photochemistry	Measures of symbiont photochemistry, chromoprotein content
	Growth	Coral or corallite growth measures including calcification rates, buoyant weight change, larval areal expansion, linear extension, and new growth branches
	Nutrient content	Total protein or carbohydrate content present in hosts or whole holobiont tissues
	Bleaching	Symbiont cell densities or change in cell densities, bleaching index scores (a proxy for symbiont cell density), and Chlorophyll A content (correlated to symbiont cell density)
	Morphology	Static intraspecific corallite measurements and larval volumes upon birth

	Symbiont community	Symbiont community indices (Leinster and Cobbold's <i>D</i>) and proportion of symbionts that are more the thermally-tolerant species (<i>Durussdinium</i> spp.)
	Immune response	Catalase and phenoloxidase activity within holobiont tissues
	Survival	Measures of survival/mortality/settlement success, including counts of settlement success or survival, percent survival/mortality at the end of a fixed period, larval survival through high temperatures, or differences in survival between control and temperature treatments
	Gamete compatibility	π -value, the percent larval contribution of various sires to various dams. Excluded from meta-analysis due to the presence of only a single estimate
<i>Coral life stage</i>	larvae	Estimates for free-swimming gamete or planula larvae stages up to successful settlement
	juvenile	Estimates from post-settlement to sexually mature adult
	adult	Estimates from colonies after sexual maturity or using coral nubbins
<i>Coral growth forms</i>	branching	Arborescent form; tree-like branching extensions
	corymbose	Finger-like extensions
	massive	Ball- or boulder-shaped corals
	encrusting	Low-spreading corals often occurring on hard, rocky substrates

columnar

Upwards-growing cylindrical corals

Temperature

covariate

Difference (in +°C) between the study's reported control or ambient temperature and the heat treatment

difference

temperature

1 *Reported heritability estimates*

2 We collected a total of 95 valid heritability estimates from 19 independent studies of
3 scleractinian corals (Fig. 1). Three studies (Császár, Ralph, Frankham, Berkelmans, & van Oppen,
4 2010; Quigley, Randall, van Oppen, & Bay, 2020; Wright et al., 2019) each involving multiple trait
5 types, provide 59% of all heritability estimates (Fig. 1-left). There was an even split of studies (9:9
6 studies) examining narrow-sense (h^2) and broad-sense (H^2) heritability, with one study (Carlson et
7 al., 2011) reporting both heritability types. However, the number of raw estimates produced by each
8 study differed markedly, with more broad-sense estimates ($n = 70$) than narrow-sense ($n = 25$). The
9 studies also differed in terms of which trait type was reported, with most studies reporting only a
10 single estimate (12 studies), and the other seven studies reporting on two to six distinct trait types.
11 Survival was the most frequently studied trait type (nine studies, 15 estimates), while bleaching (six
12 studies, 10 estimates) and growth (six studies, 23 estimates) were also diversely studied. The latter
13 often included comparisons of multiple species or symbionts within the same study, resulting in a
14 large number of estimates. Most trait type estimates originated from least two independent studies
15 estimates, save for immune response (four estimates from one study) and gamete contribution (one
16 estimate). We therefore interpret the results for immune response with caution, given that they all
17 belong to the same study, and excluded the single estimate for gamete compatibility from the
18 subsequent meta-analysis. Notably, there were limited studies of the heritability of coral
19 reproduction and fecundity, and there were no heritability estimates of thermal optimum (T_{opt}),
20 measures of performance breadth (e.g., B80, B95), or critical thermal limits ($CT_{max/min}$). Only three
21 studies reported the total phenotypic variation and/or the level of additive genetic variation, which
22 would be particularly useful for calculating metrics of evolvability (Ma, Lü, Wang, & Wan, 2014;
23 Visscher et al., 2008). We therefore recommend that future studies report these estimates of
24 variation.

25 For life stage, there were 63 estimates (from eight studies) for adults, 18 estimates for
26 juveniles (from seven studies), and 14 for larvae (from five studies), with every study reporting on

27 only a single life stage save for two reporting on two different life stages (Carlon et al., 2011;
28 Quigley, Willis, & Bay, 2017). There was similar lack of overlap across heritability types (70
29 broad-sense vs. 25 narrow-sense heritability estimates across 10 vs. 10 studies, respectively), with
30 only one study having both valid broad-sense and narrow-sense heritabilities (Carlon et al., 2011).
31 Across coral growth forms, there were 61 estimates of corymbose corals (from seven studies), 21
32 estimates for massive corals (from eight studies), nine estimates for branching/arborescent (from six
33 studies), and three and one estimates for encrusting and columnar corals, respectively (each from a
34 single study). Finally, 14/19 studies (83/95 estimates) recorded temperatures, and thus the effect of
35 temperature manipulation on heritability could be examined for these studies. However, the
36 difference between the manipulated vs. control/ambient temperatures varied substantially across
37 each study, with all temperature manipulation differences being positive (i.e., control/ambient
38 conditions were always less than the treatment temperatures) but positively skewed (a few
39 experiments used temperatures that differed by 10°C between control and high temperature
40 treatments, but most used smaller elevations of temperature). For example, 29 estimates of
41 heritability originated from control/ambient conditions (from seven separate studies), while 54
42 estimates were obtained from above-ambient temperature treatments (from 12 studies).
43 Additionally, there was limited overlap of control and heated temperature differences for some trait
44 types, making it difficult to compare the effect of temperature for trait types such as symbiont
45 community (two control estimates), morphology (one control estimate) and gene expression (nine
46 temperature differences, but no control estimates).

47

48 **Meta-analysis approach**

49 We used the *R* package *metafor* (Wolfgang Viechtbauer, 2010) to fit mixed-effects meta-
50 analytic models to $\ln(h^2 + 0.2)$ transformed heritability estimates (h^2_T) and associated estimate
51 sampling variance, while accounting for both fixed and random effects. Due to some missing
52 combinations of explanatory factors within the dataset (e.g., not all traits were measured for all life

53 stages, or for all coral growth forms), the complete dataset only allowed us to consider additive
54 effects of trait type, heritability type, life stage, and growth form in an overall analysis. Temperature
55 was not controlled for or measured in all studies, and thus was excluded as a covariate at this stage.
56 To further assess the robustness of this model and examine interactions, we then analyzed subsets of
57 the complete dataset to test for: (a) trait \times life stage interactions, (b) trait \times heritability interactions,
58 and (c) main effects and interactions involving growth form. Finally, we examined a subset of the
59 complete data that reported treatment temperature differences relative to ambient temperature,
60 including trait \times temperature difference interactions and additive effects of life stage, heritability
61 type, and growth form. All models were fit using more conservative t-distribution approximations
62 of confidence intervals in the case of multi-level random effect models, and final models fit using
63 the more conservative Knapp and Hartung (2003) adjustment for single-level random effect meta-
64 models when multi-level random effects structures were not selected during model selection
65 (VanAert & Jackson, 2019; Wolfgang Viechtbauer, 2010).

66 We considered the top model for each analysis as the model with the lowest Akaike's
67 Information Criterion, corrected for small sample sizes (AICc). We considered this model a
68 substantial improvement over other candidate models when the difference in AICc scores (ΔAICc)
69 was greater than two (Burnham & Anderson, 2004). We followed the four-step model selection
70 strategy outlined in Zuur et al. (2007, 2009): (1) define the 'beyond optimal' fixed effects structure
71 – that is, the most conceivably complex yet biologically relevant fixed effects possible, (2) select
72 (via the lowest AICc value) the optimal random effects structure for models fit using restricted
73 maximum likelihood (REML), (3) select (via the lowest AICc value) the optimal fixed effects
74 structures for models fit using maximum likelihood, (4) re-fit the final model using REML. Study
75 and species were highly confounded and precluded the inclusion of both as random effects within
76 the same model. Thus, we fit models using one of the following random effects structures: estimate
77 ID only (1|estimate ID), study ID only (1|study ID), species only (1|species), estimate ID nested

78 within its respective study ID (1|study ID/estimate ID), estimate ID nested within species
79 (1|species/estimate ID), or a random effect variance fixed at zero.

80 After fitting models for each analysis, we examined the level of among-study heterogeneity
81 (τ^2) using the I^2 index (Higgins & Thompson, 2002), which provides an estimate of the among-
82 study variance relative to the total variance not explained by the fixed effects. We used the Q_E
83 statistic to test for significant residual heterogeneity after accounting for fixed effects (W.
84 Viechtbauer, López-López, Sánchez-Meca, & Marín-Martínez, 2015). Higher proportions of
85 heterogeneity indicate that variation in true effect size of heritability is a distribution of study
86 effects (i.e. due to methodological or other study differences), whereas lower heterogeneity
87 indicates that any among-study heterogeneity is likely small relative to measurement error, so
88 studies are measuring a common heritability value (Ban, Graham, & Connolly, 2014; Higgins &
89 Thompson, 2002). We report the pseudo- R^2 for meta-analytical models, computed by comparing the
90 difference in $\hat{\tau}^2$ estimated using models including fixed effects ($\hat{\tau}_{ME}^2$) vs. a model with the same
91 random-effects structure, but with no fixed effects ($\hat{\tau}_{RE}^2$): $R^2 = (\hat{\tau}_{RE}^2 - \hat{\tau}_{ME}^2) / \hat{\tau}_{RE}^2$ (Raudenbush,
92 2009). Where significant interactions were found between factors with more than two levels, we
93 employed simultaneous tests for testing multiple general linear hypotheses determined by visual
94 inspection of marginal estimates, and report p-values adjusted using the single-step method.

95 Model standardized residuals were plotted against fitted values to look for strong deviations
96 from normality, and data were simulated using the fitted model and plotted with the true data to
97 assess model performance. We assessed the presence of publication bias by plotting the model
98 residuals by their precision (inverse of standard error) to produce a funnel plot (Møller & Jennions,
99 2001). We also calculated the Rosenberg fail-safe number, which indicates if model findings are
100 robust to any apparent publication bias if the number is greater than five times the number of
101 studies plus ten (Rosenberg, 2005; Rosenthal, 1991). Finally, we used Cook's distances to
102 determine highly influential points for each model (Cook & Weisberg, 1982).

103

104

105 **High heritability of coral traits**

106 The results of the overall analysis reveal that the heritability of coral traits has considerable
107 heterogeneity that can be explained by trait type. The final selected model used trait type as the sole
108 explanatory factor, and had substantial residual heterogeneity ($QE_{85} = 478, p < 0.0001$), with the
109 total percent of variance not attributable to sampling error, $I^2_{total} = 91\%$, composed of
110 predominantly between-study variance ($I^2_{study} = 57\%$ of total), but with substantial within-study
111 variance ($I^2_{estimate} = 34\%$ of total) as well. In all models, random effects involving estimate ID, study
112 ID, or estimate ID nested in study ID were always selected, with no support for random effects
113 involving species.

114 Trait type was by far the most important predictor of heritability across all studies (Table
115 S1), with traits such as gene expression having low heritability ($h^2 < 0.25$); photochemistry, growth,
116 nutrient content, symbiont abundance, morphology, and symbiont community having moderate
117 heritability ($h^2 = 0.25\text{--}0.5$); and immune response and survival/larval settlement success having the
118 highest heritability estimates ($h^2 > 0.5$; Fig. 2). However, models that included additive effects of
119 trait type + heritability type and trait type + life stage were supported by model selection (i.e., they
120 fit almost as well as the model with trait type alone). However, the effect sizes of both were small
121 relative to the effect of different trait types (Table S2; Fig. S2). For example, broad-sense
122 heritabilities were 1.4–2.1 times higher than narrow-sense heritability and varied by a factor of 1.1
123 to 2.1 across different life stages (within the same trait type), whereas trait type differences were
124 much larger, being up to 6.7 times larger in the case of survival vs. gene expression. Estimates for
125 the mean heritability of different traits ranged from low to high, but most traits were moderately
126 heritable (Fig. 2). Gene expression traits had the lowest estimated mean heritability ($h^2=0.12$), while
127 survival had the highest ($h^2=0.79$), followed by immune response ($h^2=0.62$), with the other
128 estimated trait mean heritabilities falling between 0.26–0.50 (Fig. 2). One estimate in particular, a
129 value of 0.92 for *Acropora millepora* (Wright et al., 2019), drove the high heritability of immune

130 response (Cook's distance = 5.2), while all other Cook's distances were relatively low (< 2). Thus,
131 the estimated high heritability of immune response should be interpreted cautiously.

132 The final model's funnel plot exhibited no signs of publication bias (Fig. S1), and the fail-
133 safe number (i.e. the number of null-result studies required to overturn a significant result) was an
134 order of magnitude above five times the number of studies plus ten (1,285 >> 100), indicating that
135 the model findings are robust to any underlying publication bias.

136

137 *Heritability across trait types in other organisms*

138 Heritability differences across trait types have been widely reported in other taxa (Flood et
139 al., 2016; Mousseau & Roff, 1987; Polderman et al., 2015; Wheelwright et al., 2014). Life history
140 traits closely tied to fitness (e.g., longevity, fecundity) are often maintained due to strong stabilizing
141 selection and thus exhibit lower heritability compared to morphological, physiological, and
142 behavioural traits (Mousseau & Roff, 1987; Price & Schluter, 1991; Teplitsky et al., 2009;
143 Wheelwright et al., 2014). However, traits may also have low heritability due to a large contribution
144 to total variance by non-additive genetic variation, environmental variation, or through maternal
145 effects (the latter likely to be less pronounced in broadcast spawning corals than in organisms with
146 higher levels of parental care). In our analysis, gene expression had the lowest heritability, which is
147 consistent with many other studies noting the low heritability of mRNA (i.e., the 'missing
148 heritability' problem, Zuk et al. 2012; Yang et al. 2014). While the exact cause of missing
149 heritability for gene expression measures has yet to be determined, it may be due to highly variable
150 gene expression both within (i.e. low repeatability) and among individuals, or to epistatic gene
151 interactions, or some combination of both (Yang et al., 2014; Zuk et al., 2012). The heritability of
152 symbiont community composition was much higher than heritabilities estimated for the diversity of
153 human gut microbes ($h^2 = 0.019$), which is predominantly environmentally rather than genetically-
154 determined (Rothschild et al., 2018). However, beneficial microbes that are related to metabolic
155 health, such as gut bacteria of the family Christensenellaceae, and microbiomes of mice in

156 controlled laboratory environments (Org et al. 2015), show much higher heritabilities of $h^2=0.3-0.6$,
157 more consistent with our findings. Photochemical traits were estimated to have modest heritability
158 in our analysis; however, only two studies which included only broad-sense estimates were
159 available ($H^2 = 0.26$). In plants, broad-sense heritability of photosynthetic traits is variable but can
160 be very high (e.g., $H^2 = 0.87, 0.5 - 0.99$, and 0.99 ; Geber and Dawson 1997, Flood et al. 2016, and
161 Tuhina-Khatun et al. 2015, respectively). Moderate narrow-sense heritability estimates, similar to
162 those reported here, have been reported for narrow-sense heritability of maximum quantum yield in
163 plants ($h^2 = 0.12-0.34$) (Qu et al., 2017). Heritability associated with bleaching and symbiont
164 abundance in corals (often using chlorophyll content as a proxy) was estimated overall as $h^2 = 0.36$,
165 which is similar to estimates of broad-sense heritability of chlorophyll content in plants (e.g., $h^2 =$
166 $0.44-0.49$ in *Oryza sativa* L., Tuhina-Khatun et al. 2015).

167

168 **Life stage and heritability type, but not growth form, mediate trait heritability**

169 Using a data subset to examine trait type and life stage interactions, a model of trait type \times
170 life stage + heritability type with a random effect of estimate ID only was preferred under model
171 selection (Table S3). Other analyses of trait type versus heritability type interaction and trait type
172 and growth form interaction found further support for a trait type \times life stage interaction (see
173 Supplementary Text S3). The final meta-model had moderate levels of heterogeneity among
174 estimates ($QE_{55} = 96, p = 0.0005; I^2_{total} = 47\%$) and fixed effects helped explain much of the
175 variation in heritability estimates (pseudo $R^2 = 78\%$). Parameter estimates for all trait types were
176 similar to the previous overall model estimates (Fig. 3; Fig. S3; Table S4), but there were
177 significant interactions for growth and bleaching in juveniles relative to other life stages as well as a
178 for nutrient content in adults (Fig. 3; Table S4). Cook's distances for the trait type \times life stage +
179 heritability type model were low overall (≤ 2), but three points had moderate leverage on the
180 analysis (Cook's distance = $2.9-3.9$), but the growth:juvenile interaction term remained important
181 when any or all were excluded from the analysis. Coral growth form was never an important

182 predictor of heritability, and species was never selected as an important random effect, suggesting
183 that taxonomic differences may be too small or variable to detect, given the data currently available.

184 Life stage had a strong effect for certain trait type–heritability type combinations (Fig. 3;
185 Table S4). For example, the estimated narrow-sense h^2 for bleaching metrics in adults was 9.1 times
186 the same h^2 for juveniles, and two times the bleaching H^2 value in adults versus larvae. Growth and
187 nutrient content broad-sense heritability also differed across life stage, with adult growth H^2 being
188 3.1 times that of juveniles and nutrient content H^2 being 3.9 times greater in larvae vs. adults. In
189 contrast, the effect of heritability type was relatively weak (1.4 to 2.5–fold higher for broad-sense
190 heritabilities vs. narrow-sense when controlling for trait type and life stage) compared to the effect
191 of trait type on heritability, which was up to 13.2 times higher heritability when comparing h^2
192 between juvenile bleaching vs. survival (Fig. 3; Table S4). However, this difference in broad- vs.
193 narrow-sense heritability type indicates the presence of substantial but not overwhelming non-
194 additive genetic variation as a portion of the total genetic variation present in broad-sense
195 heritability.

196

197 *Low adaptive potential of juvenile growth and bleaching*

198 Juvenile growth was much less heritable relative to adult growth, while bleaching was less
199 heritable in juveniles relative to both larvae and adults, highlighting the differential adaptive
200 potential of coral life stage to selection for some trait types. This reduced bleaching heritability
201 from larvae to juveniles may be the result of previous strong stabilizing selection on growth and
202 bleaching traits in juveniles, thus driving reduced additive genetic variance through the fixation of
203 alleles and resulting in lower heritabilities compared to other traits (Fisher, 1930; Teplitsky et al.,
204 2009). Indeed, bleaching events likely represent a strong selective pressure for juvenile corals
205 (Dajka et al., 2019; T. P. Hughes et al., 2019). Similarly, reductions in growth may result in
206 increased mortality due to overgrowth competition and size-dependent predation (Doropoulos,
207 Ward, Marshall, Diaz-Pulido, & Mumby, 2012; Madin et al., 2014; Vermeij & Sandin, 2008). There

208 is also evidence that increases in additive genetic variance (V_A) may occur via mutation
209 accumulation across an organism's lifetime (Wilson et al. 2008b). Moreover, reduced importance of
210 local environment with age can result in reduced relative total variation, V_P (e.g., the Wilson effect,
211 Bouchard Jr. 2013). Both of these processes can therefore result in older life stages having higher
212 heritability estimates. Making the distinction between these processes requires examining changes
213 in V_A and V_P across an organism's lifetime, which no coral studies have done to date.

214 Increased disturbances related to anthropogenic climate change are likely to select for
215 different species traits and communities (Herben, Klimešová, & Chytrý, 2018; Pratchett,
216 McWilliam, & Riegl, 2020), but little is known regarding selection on life stages within the same
217 trait. With increased frequency of bleaching events resulting in more free space being made
218 available to coral recruits, the adaptive potential of juvenile coral growth rates may determine
219 which corals become predominant in future communities. However, negative trade-offs between
220 bleaching and growth have been observed for coral symbionts (Berkelmans & Van Oppen, 2006;
221 Cunning, Gillette, Capo, Galvez, & Baker, 2015; Little, Oppen, & Willis, 2004) and juvenile coral
222 hosts (Carly D. Kenkel, Almanza, & Matz, 2015; Morikawa & Palumbi, 2019), such that more
223 thermally-specialized holobionts may exhibit reduced growth rates in ambient conditions. If these
224 phenotypic trade-offs are genetically based, the genetic correlation between the two may constrain
225 their evolution to climate change and thus would explain why the estimated heritabilities for
226 juvenile bleaching and growth are lower compared to other life stages. More study of genetic
227 correlations in juveniles is required to understand how juveniles are likely to respond to selection
228 due to climate change; however, one laboratory selection experiment on adult fragments from
229 *Acropora millepora* did find a significant positive genetic correlation ($r_g = 0.19$) between bleaching
230 and growth (Wright et al., 2019). With increased study of narrow-sense heritabilities and especially
231 genetic correlations among traits and at different life stages, the constraints on corals' responses to
232 environmental change will come into sharper focus.

233

234 *Confounding sources of variation*

235 Our review of the literature highlights some potential sources of bias in heritability estimates
236 that are not well-controlled in coral studies to date. Studies that do not use shared common
237 environments may overestimate heritability by confounding environment-driven phenotypic
238 variation with additive genetic variation, such as when related individuals occur in the same
239 environment and thus acclimatize similarly. Importantly, no studies examining adult corals raised
240 corals to adulthood in a shared common environment, and thus do not control for preconditioning or
241 canalization differences among colonies (Putnam & Gates, 2015). However, coral larvae and
242 juveniles were almost always raised in shared common environments during spawning and
243 fertilization, thus larvae and juvenile heritability estimates are less likely to be overestimated due to
244 this phenomenon. Despite this, there are a number of traits with higher heritability for larvae and/or
245 juveniles relative to adults, such as survival, gene expression, nutrient content, and morphology
246 (Fig. 3). This suggests that, at least for these traits, the variation associated with preconditioning and
247 plasticity is unlikely to be particularly large relative to the additive genetic variance. Moreover,
248 visual inspection of residuals suggested no additional unexplained variation that might be associated
249 with whether or not a shared common environment was used (Fig. S9 in Supplementary Code C).

250 Our results are also affected by other sources of phenotypic variation not accounted for in
251 present studies, such as parental and epigenetic effects. Parental effects may have a larger influence
252 on heritability than previously assumed (C. D. Kenkel, Setta, & Matz, 2015; Noble, Mcfarlane,
253 Keogh, & Whiting, 2014), and may be especially important for brooding corals in which the
254 offspring develops within the parent colony as well as for species inheriting their symbiont
255 communities directly from parents (i.e. vertical transmission) (C. D. Kenkel et al., 2015; Quigley et
256 al., 2017). Vertically-transmitting brooders and broadcast spawning species make up a minority of
257 species examined (3/19 and 9/19, respectively), with the remaining being horizontally-transmitting
258 spawners. Similarly, the number of heritability estimates from vertical transmitters made up only
259 9/95 and 16/95 heritability estimates, respectively, and thus parental effects via brooding and/or

260 vertical transmission would have impacted a minority of estimates. Epigenetic effects may also
261 inflate heritability estimates (Putnam & Gates, 2015). In studies of multicellular animals, there has
262 been little support for epigenetic inheritance via CpG methylation (Torda et al., 2017), although at
263 least one recent study in corals has found such evidence (Liew et al., 2020). Further evidence is
264 needed to determine if epigenetic changes confer fitness benefits similar to additive genetic effects
265 (Torda et al., 2017), thus future studies aiming to separate phenotypic variation specific to parental
266 effects, symbiont composition, epigenome, and additive genetic effects would be especially
267 valuable.

268

269 **Manipulated temperature has negligible effect on heritability**

270 When examining only studies that controlled for temperature, the magnitude of the
271 experiment temperature difference relative to ambient or control conditions had only a marginal
272 effect on the recorded heritability estimate (Fig. 4). The temperature difference values were all
273 positive and positively skewed, thus we square-root transformed the temperature difference data in
274 order to reduce the leverage of estimates obtained from studies using these large temperature
275 differences. After subsetting the data to exclude studies that did not report the temperature treatment
276 used relative to ambient conditions, we examined whether an interaction between trait type and
277 temperature difference was supported. Model selection favored a model of trait + heritability type,
278 with some support for alternative models of trait type only and trait type \times temperature difference
279 (Table S9; $\Delta\text{AICc} = 0.50$ and 1.89 , respectively) using random effects of estimate ID nested in
280 study ID. The trait + heritability type model had similar effect sizes compared to those in our
281 analyses presented above (Fig. S6; Table S10). The effect of trait type in the trait type \times
282 temperature model (3rd-optimal model) saw heritability differences up to a factor of 3.4–4.6 when
283 the temperature was increased +1°C to +3°C above ambient. However, within the same trait type,
284 temperature alone had a reduced effect, with heritability differences between a factor of 0–2.1 and
285 0–1.6 for +1°C and +3°C, respectively. More specifically, temperatures +1°C above ambient

286 resulted in immune response heritability increasing by a factor of 2.1, while an increase of +3°C
287 above ambient would increase heritability by a factor of 1.6. This interaction was primarily driven
288 by a single estimate of immune response (Cook's distance = 7.6), and when removed, resulted in no
289 strong interactions between trait type and temperature. Within other traits, the effect of temperature
290 was even less pronounced. Bleaching traits were decreased by 28–33% for an increase in
291 temperature of +1–3°C (though this was not significant). Other traits such as growth,
292 photochemistry, and survival all showed marginal declines in heritability with increasing
293 temperature difference (~3–5% decrease in heritability with +1–3°C). Separate analyses examining
294 temperature as a categorical variable (ambient vs elevated), as well as analyses omitting ambient
295 treatments all resulted in similar weak to non-existent effects of temperature on heritability (Fig.
296 S7-8; Tables S11-14), suggesting a limited effect of manipulated temperature on heritability across
297 studies. Since a model of trait type + heritability type was preferred over the model of trait type ×
298 temperature manipulation (Table S9), and with the inclusion of heritability type precluding the
299 ability to model a trait × temperature interaction, we fit an additive model of trait type, heritability
300 type, and temperature difference to estimate the marginal effect of temperature and found evidence
301 for, at most, a very weak effect of temperature (Fig. 4).

302

303 *Trait adaptation to warming temperatures*

304 Our meta-analysis suggests that the capacity for corals to adapt to warming temperatures
305 may be relatively consistent over short periods of moderately high temperature (e.g., +1–3°C, the
306 temperature increases used in most of the studies we analyzed). However, this is contingent on the
307 assumption that coral responses to temperature conditions in the lab are similar to their responses to
308 temperatures in the field. Previous studies have found that many traits are expected to respond
309 differently to climate change (Ahrens et al., 2020), that heritability measurements may change with
310 temperature (Bubliy & Loeschcke, 2002), and that the rate of temperature increase employed in
311 each study can also affect heritability (Chown, Jumbam, Sørensen, & Terblanche, 2009). Similarly,

312 previous heritability studies in insects report trait-specific interactions with temperature (Bubliy &
313 Loeschcke, 2002; Gunay et al., 2011). Current theory suggests that more extreme environments
314 should produce increased selective pressures that may reduce heritability (Charmantier & Garant,
315 2005; Falconer & Mackay, 1996; Wilson et al., 2006). However, despite expectations based on
316 theory and empirical results like those described above, differences in the temperature gradients
317 used in each study did not predict the among-treatment differences in heritability estimates for
318 corals. Specifically, temperature had negligible effect on the estimation of trait heritability, such
319 that an increase of +1°C may increase high vs. low trait heritability by 4 to 9%, respectively, while
320 an increase of +3°C results in an increase in heritability of 7 to 16% (changes that, if real, would be
321 opposite of the predicted direction). These findings indicate that populations with sufficient genetic
322 diversity are unlikely to experience a reduction in heritability associated with warmer temperatures,
323 in turn suggesting substantial retention of the capacity to adapt in the face of ongoing temperature
324 change.

325

326 **Coral thermal performance and challenges to predicting future adaptation to climate change**

327 The absence of an effect of temperature on trait heritability observed here could reflect
328 differences among studies in the way temperature treatments were applied, and/or differences in
329 how temperature effects were statistically modelled. Many traits of organisms are non-linearly
330 related to temperature, and these relationships are captured by measuring thermal performance
331 curves (TPCs). TPCs are quantified by subjecting individuals to increasing temperatures at a
332 standardized rate while repeatedly measuring performance (Angilletta, 2009; Chown et al., 2009), to
333 identify: (1) the value of maximal performance (P_{max}), (2) the temperature at which maximum
334 performance occurs i.e., the thermal optimum (T_{opt}); (3) the performance breadth (e.g., B80, B95),
335 and, somewhat related to the latter, (4) the limits of thermal performance (e.g., CT_{max}) (Angilletta,
336 2009; Bodensteiner et al., 2020; M. L. Logan et al., 2014). Measuring limits to thermal tolerance
337 involves either static assays of survival time in a constant high temperature, such as heat

338 knockdown time (Castañeda, Romero-Soriano, Mesas, Roff, & Santos, 2019; Ma et al., 2014), or
339 dynamic assays involving gradually increasing temperature until failure, such as temperature-at-
340 death and CT_{max} (Castañeda et al., 2019; Doyle, Leberg, & Klerks, 2011).

341 The way in which temperature was modelled in each of the studies analyzed herein – and
342 consequently, which component of thermal performance was captured – is likely to affect the
343 heritability estimated. For example, studies incorporating temperature treatment as a fixed effect
344 and estimating heritability using a single model (Dixon et al., 2015; Lohr & Patterson, 2017;
345 Manzello et al., 2019; Meyer et al., 2009), or studies that calculate heritability from the difference
346 in trait values between low vs. high temperature treatments (Császár et al., 2010; Dziedzic, Elder,
347 Tavalire, & Meyer, 2019; Yetsko et al., 2020) likely estimated the heritability of thermal sensitivity
348 (i.e., how performance changes as temperature changes). Conversely, studies that used separate
349 models for low-temperature and high-temperature treatments (Kirk et al., 2018; Quigley et al.,
350 2020; Wright et al., 2019; Zhang et al., 2019) produced separate estimates of the heritability of
351 performance under the two temperatures. One inherent problem with such an approach arises if
352 there is little or no variation in the trait value for one of the treatment levels (e.g., no mortality of
353 bleaching observed under control conditions or no observable growth when corals are placed in
354 extreme heat). The absence of among-individual variation in performance in these cases means that
355 the estimated heritability will always be near zero, regardless of any underlying additive genetic
356 variation associated with the trait in question. For studies of thermotolerance, obtaining heritability
357 estimates via differenced treatment values or as a fixed treatment effect (and thereby providing
358 heritability estimates indicative of the trait's thermal sensitivity) is likely preferable, but ideally
359 future studies would characterize responses based on many temperature points along the TPC to
360 obtain CT_{max} , T_{opt} , and P_{max} .

361 In this review, we were unable to assess whether heritabilities associated with thermal
362 sensitivity in performance were different from heritabilities of performance itself (Fig. S10).
363 However, the evolution of both maximal performance and the thermal sensitivity are inherently

364 linked by the shape of the TPC (e.g., a higher peak in the TPC would result in higher trait values
365 and greater trait thermal sensitivity), and thus their relationship may be correlated (Janhunen et al.,
366 2016). For example, with the evolution of higher upper thermotolerance (e.g., increasing CT_{max}),
367 organisms may face reduced thermal performance breadth and thermal plasticity (Hoffmann et al.
368 2013; Comte and Olden 2017; Baker et al. 2018). Growth and the thermal sensitivity of growth are
369 negatively correlated for one-year-old rainbow trout (*Oncorhynchus mykiss*) at low temperatures,
370 but not at higher temperatures, thus while there is moderate heritability for both growth ($h^2 = 0.46$)
371 and thermal sensitivity of growth ($h^2 = 0.24$), selection for higher growth is predicted to result in
372 increased thermal sensitivity in future generations held at low temperatures, but unlikely to affect
373 thermal sensitivity at higher temperatures (Janhunen et al., 2016). Similar trade-offs of growth vs.
374 sensitivity have been observed as well in adult rainbow trout (Sae-Lim et al., 2015). Further
375 complicating the matter, some genetic correlations among life history traits may be temperature-
376 specific (reviewed in Sgrò and Hoffmann 2004), including cases where negative genetic
377 correlations can become positive at higher temperatures and vice-versa. Thus, coral trait evolution
378 may further be complicated by (currently unmeasured) genetic correlations across TPC metrics.

379

380 **Conclusion**

381 Our meta-analysis estimates relatively high heritability for some traits, such as survival and
382 growth. This, coupled with the fact that heritability does not appreciably decline with increasing
383 temperature manipulation, suggests the potential for coral adaptation to future conditions of weak to
384 moderate climate change. Nevertheless, potential confounding factors that could bias some of our
385 heritability estimates upwards remain to be explored, including the effects of preconditioning and
386 canalization in adults, parental and symbiont effects, and transgenerational inheritance of CgP
387 methylation. Recent evolutionary models of corals consider the heritability of the thermal optimum
388 for corals, T_{opt} , to be anywhere from negligible (e.g., $h^2 = 0.01$) to low/medium (e.g., $h^2 = 0.16$ –
389 0.50) (Cropp & Norbury, 2020; C. A. Logan et al., 2021; Matz et al., 2020). However, there are no

390 available estimates for coral thermal performance traits such as T_{opt} , CT_{min} , CT_{max} , and $B80$, and our
391 knowledge of how TPC parameters co-evolve remains very limited. For example, the evolution of
392 higher thermal optima (T_{opt}) may result in reduced maximal performance (P_{max}) or performance
393 breadth (CT_{min} , CT_{max} , and $B80$). Other genetic trade-offs such as growth vs. thermotolerance for
394 both corals and symbionts may exist, further constraining coral evolution to climate change. Future
395 studies would ideally construct TPCs using multiple temperatures across a known pedigree of
396 individuals in order to calculate heritabilities and associated trade-offs for TPC parameters across
397 one or multiple traits. Combined with our current knowledge of trait heritabilities, this would allow
398 better predictions regarding thermal evolution of corals in response to climate change. Nevertheless,
399 our findings suggest that corals may be capable of adapting more rapidly to the thermal challenges
400 imposed by climate change than previously thought.

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408

409 **Data accessibility statement**

410 All extracted heritability estimates and supplementary code are available at
411 <https://github.com/ecology/heritability-meta>.

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771

772 **Figure Captions**

773

774 **Fig. 1.** Heritability estimates ($N = 95$) of various traits across 19 studies of reef-building corals.

775 Colour indicates the specific trait type (hue) and heritability type (broad-sense H^2 as lighter tint

776 circles, narrow-sense h^2 as darker shade). Left: Number of estimates reported in each study. Right:
777 Point estimates of heritability and their associated 95% confidence/credible intervals (whiskers) on
778 a logarithmic (\ln) scale. Heritability estimates closer to one indicate higher heritability and thus the
779 potential for higher rates of trait adaptation within the population. Dashed lines represent
780 heritability estimates where standard errors/confidence intervals were imputed.

781

782 **Fig. 2.** Heritability estimates \pm SE for the trait type-only model, not accounting for differences due
783 to (i.e., pooled across) life stage and heritability type. Traits are sorted along the spectrum according
784 to their overall relative heritability, with heritability closer to one indicating more heritable traits.
785 The number of estimates included in the meta-analysis for each trait type are indicated below each
786 error bar in grey. The gamete compatibility trait type is excluded due to its reliance on only a single
787 study/estimate.

788

789 **Fig. 3.** Heritability estimates \pm SE across trait types with multiple life stages (x-axis) and different
790 heritability types (lighter points: broad-sense heritability; darker points: narrow-sense heritability).
791 Associated sample sizes (number of original estimates) are adjacent to each point in grey.

792

793 **Fig. 4.** Heritability vs. study temperature difference (treatment temperature relative to
794 ambient/control temperature) for each trait type and heritability type, with the size of each point
795 represents its relative precision. Dashed lines indicate the estimated marginal mean effect of
796 temperature difference, while accounting for trait type and heritability type effects. One square-root
797 degree difference ($+1\sqrt{^\circ\text{C}}$) translates to a mean increase in $\ln[h^2+0.2]$ heritability of 0.03 ± 0.05 SE.

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