

1 **A selection-based framework for evaluating transgenerational phenotypic**  
2 **plasticity under environmental change**

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15

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18

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31 **Abstract**

32 Rapid environmental change has intensified interest in inter- and transgenerational phenotypic  
33 plasticity (TGP) as a potential mechanism by which populations may buffer the fitness  
34 consequences of climate change. Parental environments frequently induce shifts in offspring  
35 phenotypes, and such responses are often interpreted as adaptive, implicitly assumed to  
36 enhance offspring fitness and thus population persistence. However, evidence for phenotypic  
37 change alone is insufficient to demonstrate adaptive value, particularly when fitness is inferred  
38 from performance proxies rather than survival or reproduction. Here, we argue that a more  
39 agnostic null hypothesis is required when evaluating the fitness consequences of TGP: that  
40 intergenerational phenotypic shifts may be neutral, maladaptive, or context-dependent, rather  
41 than inherently beneficial. We propose selection analysis as a general, quantitative framework  
42 for assessing whether phenotypic variation associated with parental environments align with  
43 fitness variation in offspring. This approach does not seek to identify the mechanistic basis of  
44 TGP, nor to disentangle plasticity from genetic or epigenetic causation, but instead provides a  
45 diagnostic test of whether observed phenotypic shifts are associated with positive, negative, or  
46 absent selection. Using worked examples across taxa and environmental contexts, we  
47 demonstrate that apparent TGP frequently fails to align with selection, despite pronounced  
48 changes in offspring phenotype. These findings highlight the risk of over-interpreting  
49 intergenerational phenotypic responses as evidence of resilience to environmental change.  
50 Integrating selection analysis into studies of TGP provides a tractable and comparable  
51 framework for evaluating when intergenerational responses are likely to enhance, constrain, or  
52 have no effect on population persistence under environmental change.

## 53 **Introduction**

54 Climate variability is one of the most ubiquitous forces shaping ecosystems. Natural  
55 environments change at different time scales, from variation at geological scales to the  
56 immediate impacts of weather events. These different rates and magnitudes of change impact  
57 the physiology that underlies reproduction, growth, and dispersal of individuals (Baduel et al.,  
58 2024; Lema, 2020; Schulte et al., 2011), and thereby affect population dynamics and  
59 communities (Amarasekare, 2024; Lema et al., 2024). For example, the composition of  
60 communities changes in response to current climate warming as more susceptible species are  
61 replaced by species more tolerant to warmer temperature (Khaliq et al., 2024). Similarly,  
62 drought and storm activity were associated with species turn-over in a subtropical forest (Wu et  
63 al., 2024). Environmental variation, and anthropogenic climate change in particular, is causing  
64 population declines in some species. For example, declines in arctic auks (*Alle alle*) are  
65 associated with climate change-induced increases in sea surface temperatures and decreases  
66 in sea ice (Jakubas et al., 2024). Population declines may be associated with loss of genetic  
67 diversity and thereby further reduce the potential for genetic adaptation to climate change  
68 (Aagaard et al., 2022; Zhou et al., 2024). Understanding how organisms respond across  
69 generations to such rapid environmental change is therefore central to predicting population  
70 persistence under global change.

71         Many species have evolved mechanisms to respond to environmental change and to  
72 reduce or alleviate its potentially negative impacts (McGaughan et al., 2021; Seebacher et al.,  
73 2015). Darwinian selection drives genetic adaptation to favour phenotypes that perform best  
74 under prevailing environmental conditions and thereby alters population allele frequencies.  
75 However, genetic adaptation is slow relative to the potential pace of environmental change.  
76 Even rapid evolution takes tens of generations before there are significant phenotypic effects  
77 that can render a population more resistant to environmental change (Lescak et al., 2015). The  
78 pace of adaptation is therefore too slow to mount effective responses to climate change, except  
79 for species with very short generation times.

80         In addition to genetic adaptation, most if not all organisms have evolved the capacity for  
81 phenotypic plasticity, that is the expression of different phenotypes from a single genotype  
82 (Aliaga et al., 2019; DeWitt & Scheiner, 2004). Plastic responses are much faster than genetic  
83 adaptation, and therefore can be more effective in buffering individuals and populations from  
84 rapid environmental variability. Inter- and trans-generational phenotypic plasticity describes  
85 cases where parental environments influence offspring phenotypes, regardless of the  
86 underlying regulatory mechanism (Brass et al., 2021; Chapelle & Silvestre, 2022; Klughammer et

87 al., 2023; Lynch & Kemp, 2014; Stajic & Jansen, 2021). Such effects are often associated with  
88 epigenetic regulation, including DNA methylation and histone modification (Best et al., 2018;  
89 Fellous et al., 2022; Heckwolf et al., 2020, 2020; Loughland et al., 2021; Sävilammi et al., 2021;  
90 Skvortsova et al., 2018; Venney et al., 2022), but their mechanistic basis is frequently  
91 unresolved, particularly in non-model organisms.

92 Inter- and trans-generational transmissions of environmentally-induced information are  
93 often proposed to enable phenotypes to better match prevailing environmental conditions  
94 across single or multiple generations, and may influence population dynamics and responses  
95 to environmental change (Brass et al., 2021; Chapelle & Silvestre, 2022; Klughammer et al.,  
96 2023; Lynch & Kemp, 2014; Stajic & Jansen, 2021). As such, trans-generational plasticity has  
97 received increasing attention as a potential mechanism by which populations may persist  
98 under rapid climate change. However, the presence of intergenerational phenotypic responses  
99 alone does not provide evidence that such responses enhance fitness or population  
100 persistence. Despite their potential to improve resistance and resilience to environmental  
101 change, intergenerational phenotypic effects not necessarily beneficial. Parental environments  
102 may induce offspring phenotypes that are neutral or detrimental if they do not match the  
103 environment experienced by offspring (Beaman et al., 2016; DeWitt & Scheiner, 2004; Gibert et  
104 al., 2019; Schwanz et al., 2020). Moreover, changes in regulatory states can arise  
105 stochastically, independent of environmental cues. For example, spontaneous epimutations  
106 and age-related changes in DNA methylation can accumulate over time (Bertucci-Richter et al.,  
107 2024; Bogan & Yi, 2024; Han, 2024; Johannes & Schmitz, 2019), generating phenotypic variation  
108 that may have neutral or negative fitness consequences.

109 In a now seminal paper, Gould and Lewontin (1979) argued against the "adaptationist  
110 program" that accepted the "near omnipotence of natural selection in forging organic design  
111 and fashioning the best among possible worlds". In other words, they observed that studies  
112 often implicitly assumed that observed biological traits were the result of selection optimising  
113 fitness. However, as Gould and Lewontin (1979) point out, there are a myriad stochastic  
114 processes, constraints, and historical contingencies that can produce patterns that are not  
115 adaptive (Nielsen, 2009). A similar adaptationist bias can arise in studies of trans-generational  
116 plasticity. Environmentally-induced intergenerational phenotypic shifts are frequently  
117 interpreted as adaptive responses to environmental change without explicit tests of fitness  
118 consequences. Empirical evidence that such responses enhance fitness under climate change  
119 remains limited and context dependent (Byrne et al., 2020; Gilbert & Warner, 2026).

120 Here we propose selection analysis as a quantitative framework to evaluate whether  
121 phenotypic variation associated with inter- and trans-generational plasticity (hereon referred to  
122 collectively as ‘TGP’) aligns with variation in offspring fitness. These methods, grounded in  
123 quantitative genetics, quantify the direction and strength of selection acting on phenotypic  
124 traits, independent of their genetic or epigenetic basis. Rather than testing whether TGP evolved  
125 adaptively or identifying its mechanistic origins, selection analysis provides a diagnostic tool to  
126 assess whether intergenerational phenotypic shifts are beneficial, neutral, or detrimental with  
127 respect to offspring fitness. The goal of incorporating this approach is to generate formal,  
128 comparable estimates of selection across systems, thereby improving inference about the  
129 fitness consequences of TGP under environmental change.

130

### 131 **Evaluating the fitness consequences of intergenerational phenotypic plasticity**

132 There is a growing focus on the potential benefits of “adaptive” TGP as a mechanism by which  
133 populations may persist in the face of rapid environmental change (Galloway & Etterson, 2007;  
134 Harmon & Pfennig, 2021; Via et al., 1995). Rather than relying on selection on standing genetic  
135 variation, which typically produces population-level responses only after many generations,  
136 parental environments may directly influence offspring phenotypes via regulatory mechanisms  
137 that operate across generations (Burggren, 2015; Skvortsova et al., 2018). Such responses are  
138 often interpreted as adaptive because they may increase offspring performance under  
139 environmental conditions similar to those experienced by parents, and buy time for evolution  
140 via genetic adaptation to take place (Chevin et al., 2010). While the utility of TGP is appealing  
141 particularly in the face of anthropogenic effects, this interpretation rests on the assumption that  
142 environmentally-induced phenotypic shifts are aligned with fitness benefits, an assumption that  
143 is rarely explicitly tested (Byrne et al., 2020; Gilbert & Warner, 2026; Sánchez-Tójar et al., 2020).

144 We suggest that a more agnostic approach is required to evaluate the fitness  
145 implications of TGP. Importantly, intergenerational phenotypic effects are not always beneficial  
146 or “adaptive” (DeWitt et al., 1998). For example, moderate heat stress experienced by parents  
147 can induce offspring phenotypes that confer increased tolerance to elevated temperatures,  
148 whereas more severe or prolonged stress can disrupt cellular function and induce DNA  
149 methylation patterns that ultimately reduce offspring performance and survival (Murray et al.,  
150 2022). In such cases, the direction and magnitude of parental environmental effects determine  
151 whether offspring phenotypes are beneficial, neutral, or detrimental with respect to fitness.

152 Even when TGP is associated with changes in offspring performance, such measures are  
153 often insufficient to conclude that intergenerational responses are adaptive. Logistical

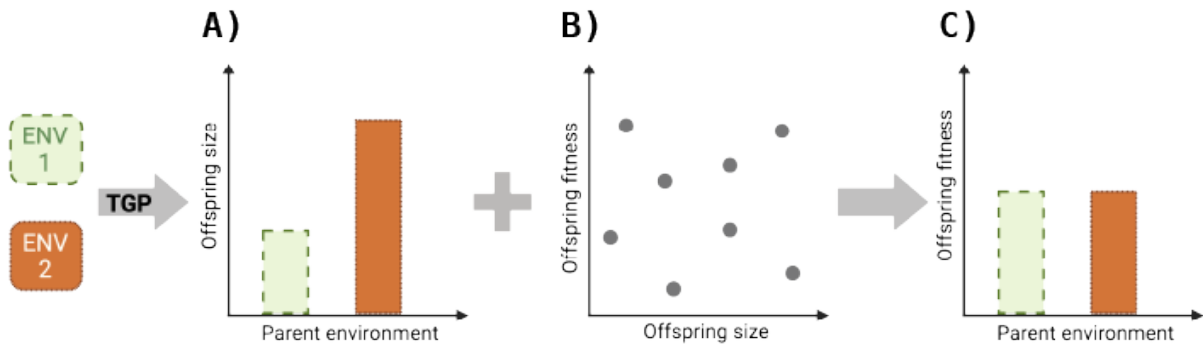
154 constraints often prevent direct measurement of survival or reproductive output, particularly in  
155 long-lived or field-based systems. As a result, the literature is dominated by proxies of fitness,  
156 including offspring size, growth rate, metabolic capacity, and locomotor performance  
157 (Donelson et al., 2012; Le Roy et al., 2017; Pettersen et al., 2019). While such traits are  
158 biologically meaningful, they may be weakly correlated with fitness, or even trade off against  
159 fitness components such as survival or reproduction (**Figure 1A**). For example, trade-offs  
160 between reproduction and aging or lifespan (Cohen et al., 2020), growth (Larue et al., 2021),  
161 resistance to parasitism (Ives et al., 2020), and survival (Cox et al., 2010) are well documented.  
162 Consequently, shifts in offspring phenotype alone cannot be interpreted as evidence that TGP  
163 enhances fitness or populations persistence. Rigorous tests of the adaptive nature of TGP  
164 should include estimates of offspring fitness (lifetime reproductive output) or at least major  
165 components of fitness (survival, mating success [fertility], or reproductive output) (Burgess &  
166 Marshall, 2014; Matesanz et al., 2022).

167 Conversely, TGP can act as a conduit for transmitting environmental stress across  
168 generations (Marshall & Uller, 2007). Challenging or damaging parental environments may  
169 induce offspring phenotypes that reduce fitness rather than enhance it. For example, chronic  
170 paternal stress in mice alters sperm microRNA composition and is associated with altered  
171 hypothalamic-pituitary-adrenal axis function in offspring (Rodgers *et al.* 2015). However, as with  
172 putatively adaptive responses, evidence for maladaptive TGP requires explicit demonstration  
173 that such intergenerational effects reduce offspring fitness, rather than simply altering offspring  
174 phenotype or physiology.

175 Here, we argue that an a priori assumption of neutral intergenerational phenotypic  
176 effects provides a more conservative and unbiased starting point for studies of TGP. In this  
177 context, we use the term “epigenetic drift” to describe intergenerational phenotypic variation  
178 that arises independently of selection and does not necessarily confer fitness benefits,  
179 analogous to genetic drift. Such variation may result from stochastic changes in regulatory  
180 states, incomplete transmission of environmental information, or mismatches between  
181 parental and offspring environments (Bertucci-Richter et al., 2024; Bogan & Yi, 2024; Han, 2024;  
182 Johannes & Schmitz, 2019). Importantly, this concept is not intended to imply a specific  
183 molecular mechanism, nor to suggest that all intergenerational effects are epigenetic in origin  
184 (McGuigan et al., 2021). Rather, it provides a parsimonious null hypothesis against which  
185 adaptive or maladaptive fitness consequences of TGP can be tested.

186 From this perspective, the adaptive, neutral, or detrimental nature of TGP cannot be  
187 inferred from phenotypic change alone and instead must be evaluated experimentally by

188 quantifying fitness consequences. Below, we outline how selection analysis can be integrated  
 189 into studies of inter- and trans-generational plasticity to assess whether phenotypic variation  
 190 associated with parental environments aligns with selection on offspring fitness. This approach  
 191 does not seek to disentangle the mechanistic origins of TGP or to test whether such responses  
 192 evolved via selection but rather provides a general diagnostic framework for evaluating fitness  
 193 alignment under environmental change.  
 194



195  
 196 **Figure 1.** Measures of parental performance, such as offspring size or aerobic scope may not be  
 197 correlated, or may even trade-off, with actual fitness (that has not been measured). In this  
 198 hypothetical example, there appears to be adaptive inter-generational plasticity in offspring  
 199 size: parents reared in Environment 2 ('ENV 2', orange shading) produce larger offspring than  
 200 parents exposed to Environment 1 ('ENV 1', green shading); **A**). However, if there is no  
 201 relationship between offspring fitness and size (**B**) then there will be no fitness benefit of the  
 202 inter-generational plasticity (**C**). It is therefore essential to link measures of offspring fitness  
 203 (survival and reproductive output) to plastic traits (size in this case) to avoid misinterpreting the  
 204 adaptive nature of TGP.

205

### 206 **Applying a quantitative genetics framework to TGP**

207 To evaluate whether TGP contributes to resilience to environmental change, it is necessary to  
 208 quantify whether phenotypic shifts induced by parental environments are associated with  
 209 fitness consequences in offspring. Specifically, TGP can be considered beneficial when it aligns  
 210 with selection (positive covariation with survival or reproductive output), detrimental when it is  
 211 misaligned, and neutral when there is no detectable association between phenotype and  
 212 fitness (Pettersen et al., 2024). Importantly, we use "selection" in the Lande-Arnold sense as  
 213 differential fitness among phenotypically different individuals (Lande & Stevan J. Arnold, 1983),  
 214 without implying any particular genotype-phenotype mapping. Selection therefore manifests as  
 215 non-random differences in survival or reproductive output across phenotypes (Falconer &

216 Mackay, 1996; Fisher, 1930; Haldane, 1954). While selection acts on phenotypes, its longer-  
217 term evolutionary consequences depend on inheritance and constraints (Arnold & Wade, 1984).  
218 Here, our focus is on the immediate inferential step that is often missing in studies of TGP:  
219 whether observed intergenerational phenotypic responses align with fitness variation in  
220 offspring under environmental change.

221 Phenotypic plasticity is sometimes viewed as reducing the efficacy of selection by  
222 increasing phenotypic variation without corresponding heritable variation, thereby weakening  
223 evolutionary responses (Chevin et al., 2010; Gienapp et al., 2007; Merilä & Hendry, 2014; Oostra  
224 et al., 2018). Yet, selection acts on phenotypes regardless of their underlying causes (Lande &  
225 Stevan J. Arnold, 1983). Selection analysis therefore provides a tractable and comparable  
226 approach for estimating the direction, form, and strength of associations between phenotypic  
227 variation and fitness across environmental contexts (**Box 1**). In the context of TGP, this shifts  
228 interpretation away from assuming that intergenerational traits changes indicate adaptive  
229 “buffering”, and towards explicitly testing whether those changes predict higher (or lower)  
230 fitness.

231 Selection analysis is a simple, yet powerful framework for producing standardised,  
232 comparable estimates for the relationship between phenotypes and fitness. Although it cannot  
233 distinguish phenotypic selection from an evolutionary (genetic) response to selection (Haldane,  
234 1954), it is well suited as a diagnostic tool for evaluating whether intergenerational phenotypic  
235 shifts are associated with positive, negative, or absent fitness consequences in offspring among  
236 parental treatments (**Box 1**). Related approaches have been applied to study within-generation  
237 phenotypic plasticity, for example by regressing plasticity on individual fitness (DeWitt, 1998) or  
238 genotype-mean fitness (Arnold et al., 2019). However, equivalent worked examples for  
239 evaluating TGP in explicitly intergenerational designs are rare, despite the central role such  
240 studies play in climate-change biology. Here, we provide practical guidance and worked  
241 examples to show how selection analysis can be incorporated into common parent x offspring  
242 experimental designs to generate comparable estimates of fitness alignment across  
243 environmental treatments and taxa. Selection analysis can thereby improve inference about  
244 when TGP is likely to enhance, constrain, or have no effect on population persistence under  
245 environmental change.

## 246 **Selection analysis methods**

247 Selection analysis required paired measurements of individual phenotypes and fitness within a  
248 population (Lande & Stevan J. Arnold, 1983). Fitness components such as survival, reproductive  
249 success, or reproductive output can therefore be used to quantify viability, fertility, or fecundity  
250 selection, respectively. Data collected in inter- and transgenerational studies are particularly  
251 well suited to this framework. In many experience, parents are exposed to contrasting  
252 environmental treatments and offspring are then tracked to measure variation in traits and  
253 performance. This experimental design can produce the phenotype-fitness data needed to  
254 estimate selection gradients within each combination of parental and offspring environments.  
255 However, several methodological considerations need to be addressed to isolate the effects of  
256 parental and offspring environments (see below). Once phenotypic and fitness data have been  
257 collected for each environmental combination, the selection landscape can be quantified using  
258 the following steps.

### 259 **1. Check for multicollinearity among traits.**

260 Before including multiple traits in a selection analysis, it is important to assess whether  
261 these traits are too strongly correlated with one another. If traits are highly correlated, it  
262 can be difficult to determine statistically how selection acts on each trait  
263 independently. To evaluate this, variance inflation factor (VIF) values should be  
264 calculated (Zuur et al., 2010). VIF values provide a measure of how much the variance of  
265 one trait is inflated by its relationship with other traits.

266 • A VIF value  $< 5$  generally indicates low multicollinearity, meaning the traits are not  
267 overly correlated and can be analysed together.

268 • A VIF value  $> 5$  suggests high multicollinearity, meaning it may be difficult to estimate  
269 independent selection gradients for each trait.

270 In cases of high multicollinearity, it may be necessary to i) combine correlated traits into  
271 composite variables, ii) remove one of the correlated traits, or iii) apply methods  
272 designed to handle multicollinearity such as ridge regression or principal component  
273 analysis.

### 274 **2. Define the environments in which selection will be estimated.**

275 Selection analyses are typically conducted within the ecological contexts in which  
276 individuals experience selection. These contexts may represent experimental  
277 treatments, environments, or combinations of both. For example, in the zebrafish case  
278 study presented here (Pettersen et al., 2024), selection gradients were estimated  
279 separately within each parental temperature  $\times$  parental food  $\times$  offspring temperature

280 combination. Estimating selection within environments allows the strength and  
281 direction of selection to be compared across ecological contexts.

282 **3. Choose an appropriate fitness measure.**

283 Selection analyses require a measure of individual fitness or a proxy for fitness.

284 Depending on the study system, this may include survival, reproductive output, fertility,  
285 or growth to a critical life-history stage. The statistical model used will depend on the  
286 distribution of the fitness variable. For example:

287 • Gaussian models are appropriate for continuous fitness measures such as  
288 reproductive output.

289 • Binomial or logistic models are appropriate for binary fitness measures such as  
290 survival or fertility.

291 In the zebrafish case study (**Box 1**), offspring fitness was measured as survival to two  
292 weeks post-hatching, which was analysed using logistic regression.

293 **4. Standardise trait values within environments.**

294 To ensure that selection coefficients are comparable across traits, phenotypic traits  
295 should be standardised to have a mean of zero and a standard deviation of one. This  
296 procedure places all traits on the same scale and allows the resulting coefficients to be  
297 interpreted as selection acting on one standard deviation change in the trait (Lande &  
298 Stevan J Arnold, 1983).

299 When selection is estimated within treatments or environments, traits should be  
300 standardised within each environment, rather than across the entire dataset. This  
301 ensures that trait variation is evaluated relative to the conditions in which selection  
302 occurs.

303 **5. Estimate directional selection gradients using multiple regression.**

304 Directional selection gradients can be estimated using multiple regression models in  
305 which fitness is regressed on the standardised traits of interest. Including multiple traits  
306 in the same model allows the selection gradient for each trait to be estimated while  
307 accounting for covariance among traits.

308 For example, the following logistic regression model can be used when fitness is  
309 measured as survival:  $\text{Fitness} \sim \text{Trait}_1 + \text{Trait}_2$

310 In the zebrafish case study (**Box 1**), survival was modelled as a function of standardised  
311 development time and growth rate within each environment.

312 **6. Transform logistic regression slopes to selection gradients.**

313 When fitness is measured as a binary variable (e.g., survival), logistic regression

314 coefficients are expressed on the logit scale and therefore cannot be interpreted  
315 directly as selection gradients. To obtain average directional selection gradients ( $\beta^*$ ),  
316 these slopes should be transformed using the method described by Janzen and Stern  
317 (1998). This transformation rescales logistic slopes to reflect the expected change in  
318 relative fitness associated with variation in the trait.

319 The transformation incorporates:

- 320 • the predicted survival probabilities from the fitted model, and
- 321 • the mean fitness within the environment.

322 Applying this transformation yields directional selection gradients that are directly  
323 comparable across traits and environments.

#### 324 7. Interpret the direction and strength of selection.

325 Directional selection gradients indicate whether higher or lower trait values are  
326 associated with increased fitness.

- 327 • Positive  $\beta^*$  indicates selection favouring higher trait values.
- 328 • Negative  $\beta^*$  indicates selection favouring lower trait values.
- 329 • Values overlapping zero indicate weak or absent directional selection.

330 Comparing these selection gradients with observed intergenerational shifts in trait  
331 values allows researchers to determine whether plastic changes align with, oppose, or  
332 are unrelated to the direction of selection.

#### 333 8. Visualise and interpret selection patterns.

334 Selection gradients can be visualised using forest plots or similar graphical approaches  
335 that display the magnitude and uncertainty of  $\beta^*$  estimates across environments. These  
336 visualisations facilitate comparison of the strength and direction of selection across  
337 treatments and traits. When combined with estimates of intergenerational trait shifts,  
338 these plots provide a clear framework for evaluating whether intergenerational plasticity  
339 aligns with selection on fitness.

340

#### 341 Selection on TGP is often context dependent

342 In natural populations, environments can be stressful, resources can be limited, or predation  
343 may be present, such that variation in phenotypes can pose important fitness consequences  
344 (Reznick et al., 2002). The selection analysis we describe is not explicit in identifying the  
345 mechanisms underlying changes in offspring phenotype and fitness, and nor was it designed to  
346 do so. TGP could be the result of genetic or epigenetic mechanisms. Genetic changes (i.e.,  
347 changes in allele frequencies affecting phenotypes) are relatively unlikely across single or few

348 generations unless selection pressures are very high, such as in antibiotic exposure of bacterial  
349 populations, for example (Bakkeren et al., 2020). As far as we know, the fastest explicitly  
350 demonstrated genetic adaptation to sudden environmental change in a vertebrate animal – the  
351 relatively short-lived stickleback fish (*Gasterosteus aculeatus*), took around 50 years (or ~25  
352 generations) (Lescak et al., 2015). Hence, at least among vertebrates, epigenetic mechanisms  
353 are more likely to cause TPG. To test for particular mechanisms explicitly, candidate epigenetic  
354 or genetic mechanisms would need to be manipulated experimentally, such as in a CRISPR  
355 knock-out of DNA methyltransferase 3 - the principal enzyme laying down new methylation  
356 patterns in response to environmental inputs (Loughland et al., 2021). Fitness consequences of  
357 knock-out vs wild type individuals can then be measured (Seebacher & Bamford, 2024) and  
358 analysed with the selection analysis we propose here. Such an approach would yield interesting  
359 insights, but those would not be essential to realise the benefits of the selection analysis we  
360 propose.

361

#### 362 **Key limitations and methodological considerations**

363 Selection analysis requires trait and fitness data (with sufficient variation) for individuals within  
364 each treatment. It is often logistically difficult to measure relative fitness, hence many studies  
365 obtain proxies for fitness, such as body size, growth rate, or locomotor performance  
366 (Stinchcombe et al., 2017), which can also be used in selection analysis in place of measures of  
367 reproduction and survival. For example, it is possible to analyse TGP using morphometric  
368 parameters to assess the benefits or costs to locomotor performance and hence broaden the  
369 applicability of standardised measures of selection. However, caution should be exercised  
370 when analysing TGP using fitness proxies to infer its adaptive potential (Acasuso-Rivero et al.,  
371 2019). Franklin and Morrissey (2017) found that using performance measures in place of  
372 fitness often underestimates the actual strength of selection. Given the requirement of  
373 selection analyses for datasets containing measures of many individuals often tracked  
374 longitudinally, it might be assumed that selection analyses are constrained to model organisms  
375 measured in the laboratory. Yet, selection analyses have been performed on natural  
376 populations of plants (Munguía-Rosas et al., 2011), birds (Charmantier et al., 2004), lizards  
377 (Losos et al., 2004), fish (Brooks & Endler, 2001), and mammals (Boratyński & Koteja, 2010;  
378 Kruuk et al., 2002) for several decades. Further selection estimates, particularly on TGP, are  
379 crucially needed to build an understanding of the fitness consequences of trait variation under  
380 rapid environmental change (Arnold et al., 2019; Rivkin et al., 2019; Svensson, 2023). Here we

381 outline some potential limitations and methodological considerations when utilising selection  
382 analysis for inferring the adaptive nature of TGP:

- 383 1. Parent and offspring treatments need to be applied independently wherever possible.  
384 Studying species with short brooding times, such as egg laying taxa, often allows for  
385 selection on offspring phenotypes to be analysed without the issue of parental  
386 environment as a confounding factor. For species with high parental care, however, it  
387 may be difficult to disentangle selection on offspring phenotype independently of parent  
388 environment or condition (Kielland et al., 2017).
- 389 2. Disentangling genetic versus environmental (epigenetic) effects on offspring phenotype  
390 and fitness. Most TGP studies (including the case studies examined here) test for  
391 adaptive TGP using polymorphic populations, where the underlying genetic basis of the  
392 phenotypic response to a parental stressor/environment is unknown. There is therefore  
393 a potential role for the genetic control of phenotypic plasticity, whereby differential  
394 mortality or reproductive output in offspring can be the result of the genotype in a given  
395 parental environment, to bias estimates of TGP (i.e., phenotypic change solely due to  
396 the parental environment). The conflation of within-generation selection with plasticity  
397 has been demonstrated by Santos et. al., (Santos et al., 2019) and may apply for TGP  
398 estimates. To clearly demonstrate that phenotypic changes occur in the absence of  
399 genetic differences, i.e., due to epigenetics, studies are needed that either use  
400 populations with a known genetic basis of the phenotypic response (e.g., offspring  
401 mass, metabolism, or immune response) such as via monitoring changes in allele  
402 frequencies across parental environment (e.g., temperature, drought, or parasite).  
403 Alternatively, this issue can be avoided altogether by using a protocol with either clonal  
404 replicates or isogenic lines (Signor, 2020).
- 405 3. Detecting selection requires that there is sufficient variation in both the traits of interest  
406 and fitness. One potential issue for using this type of analysis is that TGP might result in  
407 depletion of phenotypic variation in a given environment, such that even if TGP in  
408 response to a change in environment increases offspring fitness, it cannot be detected.  
409 For example, parents exposed to predation might increase offspring size, leading to  
410 relatively higher offspring survival in large versus small offspring, but selection  
411 (correlation between fitness and trait) cannot be detected statistically (Engqvist &  
412 Reinhold, 2016). In line with this, studies on populations already exposed to strong past  
413 selection, may not possess the available phenotypic variation required for selection to  
414 act on in the present generation.

- 415 4. Caution should be exercised when identifying the target of selection. For example,  
416 rather than acting upon the measured trait(s), selection may be targeting other  
417 unmeasured trait(s) that are correlated with trait(s) of interest, leading to a spurious  
418 correlation between fitness and the measured trait(s). It is important therefore to  
419 carefully select trait measurements, and to measure and test for selection on multiple  
420 traits.
- 421 5. It is important to apply a treatment independently from other confounding  
422 environmental effects (Rausher, 1992; Stinchcombe et al., 2002). For example, the  
423 spatial variation of treatments can be randomised to ensure that environmental  
424 conditions other than the treatment do not systematically alter phenotypic or fitness  
425 variation.
- 426 6. Measures of lifetime reproductive output in offspring and grand offspring from TGP  
427 studies are exceedingly rare. Hence, it is difficult to determine how long the benefits or  
428 costs of epigenetic effects can persist. Future studies measuring whether fitness  
429 consequences persist throughout the life history and across generations are needed to  
430 determine the temporal stability of selection acting on TGP (Siepielski et al., 2009).  
431 These studies could also shed light on the mechanisms underlying beneficial,  
432 detrimental, or neutral TGP.
- 433 7. It is important to consider that fitness benefits for individual offspring are not always  
434 beneficial for parents (Einum & Fleming, 2000; Stearns, 1992). Selection acting on  
435 offspring traits may not lead to adaptive evolution if there is antagonistic selection  
436 acting across generations (Wilson et al., 2005). To determine whether TGP is likely to  
437 evolve in response to selection, future studies that span generations and quantify  
438 phenotypic and fitness covariances between parents and offspring, are an important  
439 next step (Wolf & Wade, 2001).

440

#### 441 **Conclusion**

442 Inter- and transgenerational phenotypic plasticity is often invoked as a mechanism of resilience  
443 to rapid environmental change. Yet phenotypic shifts alone, especially when interpreted through  
444 performance proxies, do not demonstrate whether transgenerational responses enhance  
445 offspring fitness or population persistence. Here we propose selection analysis as a diagnostic  
446 framework for evaluating the adaptive value of transgenerational responses by testing whether  
447 parental-environment-associated trait variation aligns with selection to enhance survival or  
448 reproduction. Adopting fitness-based tests of alignment between plasticity and selection will

449 sharpen prediction of population responses under environmental change, and clarify when  
450 transgenerational effects are likely to enhance, constrain, or have no effect on population  
451 persistence.

452 **BOX 1. Testing whether intergenerational shifts in development and growth are adaptive**

453 Inter- or trans-generational phenotypic plasticity (TGP) is frequently inferred to be adaptive  
454 when parental environments induce shifts in offspring traits linked to performance. In zebrafish  
455 (*Danio rerio*), development time and growth rate are commonly used as proxies for early-life  
456 success, and both traits are sensitive to parental thermal and nutritional environment  
457 (Pettersen et al., 2024). Here, we demonstrate how incorporating a simple selection analysis  
458 changes interpretation of TGP in these traits.

459

460 *Intergenerational shifts in offspring traits*

461 Across parental temperature treatments (24 °C and 30 °C), low parental food consistently  
462 produced offspring that developed more slowly (longer time to hatching) and exhibited reduced  
463 early growth rates, relative to offspring from high-food parents (**Table 1**). These shifts were  
464 observed at both offspring rearing temperatures (24 °C and 30 °C), indicating clear evidence of  
465 TGP. Without explicit fitness data, such shifts towards slower development and reduced growth  
466 might be interpreted as i) maladaptive consequences of parental stress, or ii) protective  
467 adjustments that enhance survival under stressful conditions. However, trait shifts alone  
468 cannot resolve this ambiguity.

469

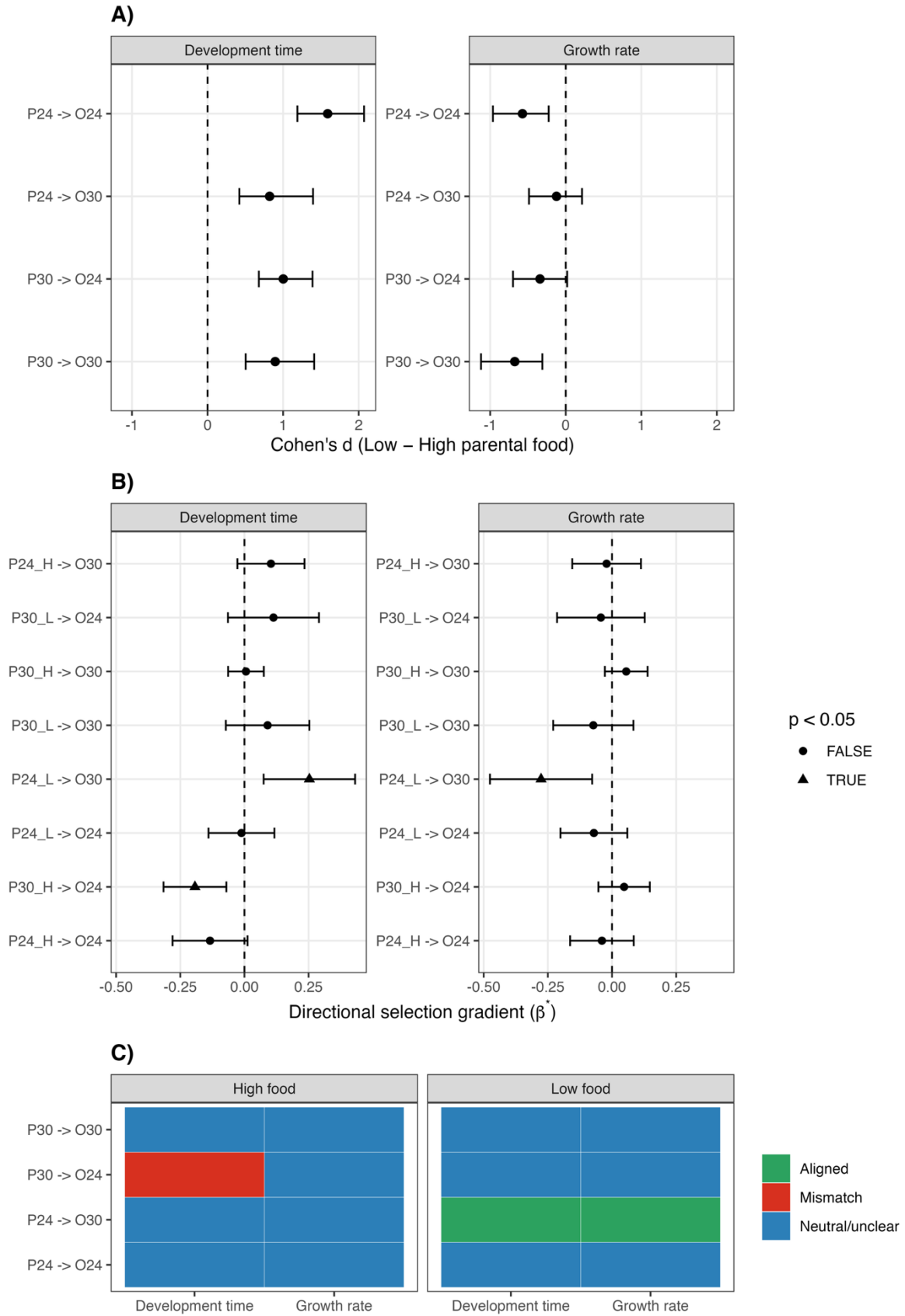
470 *Quantifying alignment of TGP with viability selection*

471 To test whether these intergenerational shifts align with fitness, we quantified viability selection  
472 by regressing relative survival (to two weeks post-hatching) on standardised development time  
473 and growth rate within each parental x offspring environment combination. Logistic regression  
474 coefficients were transformed following Janzen and Stern (1998) to yield directional selection  
475 gradients ( $\beta$ ) on a relative fitness scale. Selection gradients revealed that adaptive alignment  
476 was highly context-dependent (**Table 1**). Strong positive selection on slower development and  
477 negative selection on growth occurred only in one environmental combination (parents at 24 °C,  
478 low food; offspring at 30 °C). In this case, the direction of TGP (slower development and reduced  
479 growth under low parental food) matched the direction favoured by viability selection,  
480 consistent with beneficial IGP. In contrast, in several other environments selection was weak or  
481 absent, and in one combination (parents at 30 °C, high food; offspring at 24 °C), selection  
482 favoured faster development, opposite to the direction of the parental temperature-induced  
483 shift. Thus, intergenerational shifts that might be broadly described as adaptive based on trait  
484 patterns along were only beneficial under specific environmental conditions.

485

486 *How inference changes*

487 If analysis had stopped at documenting intergenerational trait shifts, we might conclude that  
488 parental thermal and nutritional stress modifies offspring development and growth in ways  
489 relevant to performance. However, incorporating explicit estimates of selection demonstrates  
490 that adaptive alignment is neither universal nor consistent across environments. Some trait  
491 shifts are strongly favoured by selection, others are effectively neutral, and some may oppose  
492 the direction favoured by viability (survival). This additional analytical step required only  
493 individual-level trait and survival data and does not depend on pedigree information or complex  
494 quantitative genetic modelling. Yet it fundamentally alters inference about whether TGP  
495 enhances, constrains, or has no effect on offspring fitness. Zebrafish are widely used in  
496 intergenerational and epigenetic studies, including investigations of DNA methylation dynamics  
497 (Loughland et al., 2021; Seebacher & Bamford, 2024). This case study illustrates how integrating  
498 selection analysis into such designs provides a direct and comparable test of adaptive  
499 alignment, strengthening inference about the evolutionary and ecological consequences of TGP  
500 under environmental change.



501

502

**Figure (Box 1). Testing whether intergenerational phenotypic plasticity aligns with viability**

503

**selection. A) Intergenerational shifts in offspring (“O”) development time and growth rate**

504 induced by parental (“P”) food environment, expressed as standardized mean differences  
505 (Cohen’s  $d$ ; low – high parental food) with bootstrapped 95% confidence intervals. Positive  
506 values indicate slower development or faster growth under low parental food; negative values  
507 indicate faster development or slower growth. **B)** Multivariate directional selection gradients  
508 ( $\beta^*$ ) estimated from logistic regressions of offspring (“O”) survival on standardised development  
509 time and growth rate within each parental (“P”) temperature (24 = 24 °C, 30 = 30 °C) × parental  
510 food (L = Low, H = High) × offspring temperature (24 = 24 °C, 30 = 30 °C) environment. Error bars  
511 show 95% confidence intervals. **C)** Alignment between the direction of intergenerational trait  
512 shifts and the direction of viability selection for each parental food environment. Green  
513 indicates alignment between plastic shifts and selection, red indicates mismatch, and blue  
514 indicates neutral or unclear outcomes where selection gradients overlap zero. Together, these  
515 panels illustrate how incorporating selection analysis can change inference about the adaptive  
516 value of intergenerational plasticity: although parental environment consistently altered  
517 offspring traits (panel A), only some shifts aligned with selection on survival (panels B–C). This  
518 workflow demonstrates a practical approach for testing whether intergenerational plasticity  
519 enhances offspring fitness under environmental change.

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