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1 **Prepared as a Short Communication**

2 **Low Repeatability of Aversive Learning in Zebrafish (*Danio***
3 ***rerio*)**

4
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22 Investigation, Writing – original draft preparation, Writing – review and editing,

23 Visualisation. **Susanna Zajitschek:** conceptualisation, methodology, Data Curation, Writing

24 – review and editing, Supervision. **Hamza Anwer:** Writing – review and editing. **Rose**
25 **O’Dea:** Data Curation, Writing – review and editing, Visualisation. **Daniel Hesselton:**
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32 **Summary Statement**

33 Zebrafish were trained to avoid colour stimuli through negative associations with a mild
34 electric shock. Very low repeatability estimates suggest individuals were not distinguishable
35 by their aversive learning abilities.

36 **Abstract**

37 Aversive learning – avoiding certain situations based on negative experiences – can
38 profoundly increase fitness in animal species, yet no studies have systematically quantified its
39 repeatability. Therefore, we assessed the repeatability of aversive learning by conditioning
40 approximately 100 zebrafish (*Danio rerio*) to avoid a colour cue associated with a mild
41 electric shock. Across eight different colour conditions zebrafish did not show consistent
42 individual differences in aversive learning ($R = 0.04$). Within conditions, when zebrafish
43 were conditioned to the same colour, blue conditioning was more repeatable than green
44 conditioning ($R = 0.15$ and $R = 0.02$). Overall, aversive learning responses of zebrafish were
45 weak and variable. We speculate that the effect of aversive learning might have been too
46 weak to quantify consistent individual differences, or directional selection might have eroded
47 additive genetic variance. We also discuss how confounded repeatability assays and
48 publication bias could have inflated estimates of repeatability in the literature.

49 **Introduction**

50 Animals use the cognitive process of learning, which can be defined as a change in behaviour
51 due to past experience, to respond to the environment (Kawecki, 2010). Learning has a
52 profound influence on survival and reproductive success (Krebs & Davies, 1987; Skinner,
53 1984), and has been studied in a wide range of taxa. For example, individual learning speed
54 has been correlated with foraging performance in bees (Raine & Chittka, 2008) and
55 grasshoppers (Pasquier & Grüter, 2016); and greater cognitive capacity has been linked to
56 higher reproductive success in magpies (Ashton et al., 2018) and male robins (Shaw et al.,
57 2019), as well as to healthier body condition in wild primates (Huebner et al., 2018).

58
59 Animals learn through association, which is reinforced differently by positive and negative
60 experiences (appetitive and aversive learning, respectively). Appetitive learning takes place
61 when individuals associate a stimulus with a ‘positive’ event, usually a food reward stimulus,
62 whereas in aversive learning the association is with a ‘negative’ event, usually a fear inducing
63 stimulus. Failing to learn from positive experiences (appetitive learning) prevents a potential
64 benefit (i.e., a minor opportunity cost). Failing to learn from negative experiences may yield
65 an immediate fatal cost. Therefore, both types of learning can increase lifetime fitness and
66 drive natural selection, but appetitive learning may be under weaker selection than aversive
67 learning.

68
69 For traits to evolve they need heritable variation that can be subject to selection. For labile
70 traits (i.e., traits expressed more than once over a lifetime) the consistency of individual
71 differences in trait expression indicates potential heritability. The common approach to
72 quantify consistent individual differences in eco-evolutionary studies is estimating the
73 statistical index ‘repeatability’ (R ; otherwise known as the ‘intra-class correlation coefficient’

74 or ICC; Lessells & Boag, 1987; Nakagawa & Schielzeth, 2010). Repeatability partitions
75 variance into within-individual (residual) and between-individual components. Biologically,
76 the repeatability of a trait indicates the amount of observed variance which is due to
77 individuals sustaining trait differences between each other (Nakagawa & Schielzeth, 2010),
78 but estimates can be inflated by measurement errors and experimental confounds (Dohm,
79 2002; Niemelä & Dingemanse, 2017).

80

81 Generally, behavioural traits are moderately repeatable ($R = 0.34$; Bell et al., 2009; cf.
82 Holtmann et al., 2017), with cognitive behavioural traits showing somewhat lower
83 repeatability ($R = 0.15 - 0.28$; Cauchoux et al. 2018). Our understanding of how natural
84 selection shapes the evolution of cognitive traits remains poor (Boogert et al., 2018). Despite
85 the extensive literature on aversive learning, no published study has comprehensively
86 quantified its repeatability (but note Cauchoux et al. (2018) includes three unpublished studies
87 with some measures of aversive learning). To reduce this knowledge gap, we quantify the
88 repeatability of aversive learning behaviour in zebrafish (*Danio rerio*), a popular model
89 organism in cognitive science (Gerlai, 2016; Norton & Bally-Cuif, 2010). Zebrafish exhibit a
90 range of distinct behaviours that can be measured in previously established assays
91 (Fangmeier et al., 2018; Meshalkina et al., 2017).

92

93 Here, we use an avoidance conditioning assay — associating a visual cue with a mild electric
94 shock (see Fig. 1A-E) — to thoroughly assess the repeatability of colour preferences and
95 aversive learning in both male and female zebrafish. We expect individuals to consistently
96 differ in their aversive learning speeds (i.e., separation of better and worse learners), but do
97 not have particular expectations for sex differences. We estimate repeatabilities in two
98 different ways. First, we examine repeatability across different colour pairs (four different

99 pairs with eight possible combinations: 8 measurements per individual; Fig. 1F). Given the
100 estimates for appetitive learning summarised in Cauchoix et al. (2018), we predict a low to
101 moderate repeatability. Second, we test whether repeatability is increased in a constant
102 learning environment by using just one colour pair (both combinations of green and blue; 3
103 repeated measurements per individual for each colour; Fig. 1F). For both types of
104 repeatability measurements (within and across learning environment), we also quantify
105 colour preferences and their repeatabilities, to give a comparator in individual differences that
106 can be compared to aversive learning.

107

108 **Materials and Methods**

109 **Zebrafish population**

110 Adult wildtype zebrafish were bred on the 24th of January 2019 (5 months old at the
111 commencement of experiments) and maintained at the Garvan Institute of Medical Research
112 in Sydney, Australia. The wildtype stock was derived from of a mixture of Tübingen long fin,
113 AB and other unidentified strains to increase genetic diversity, which had been interbred for
114 8-10 generations. Fish were housed in 3.5L Tecniplast ZebTEC tanks (maximum of 24 fish
115 per 3.5L tank) under standard laboratory conditions (~28°C; ~pH 7.5; ~1000 µs conductivity;
116 12/12h from 7:30 light/dark rotation) and fed live *Artemia salina* nauplii twice a day and
117 commercially available fish food once per day (O.range GROW-L).

118

119 We marked juvenile fish for individual identification at around 90 days post-fertilisation with
120 coloured tags (red, brown, purple, black, white, yellow, orange, pink, or green). For marking,
121 fish were anaesthetised in a tricaine solution (4.2ml of 0.4% in 100ml of system water) for 20
122 seconds before being injected with Visible Implant Elastomer tags (VIE, Northwest Marine

123 Technologies, Inc.; Shaw Island, Washington, United States). We injected fish twice (unless
124 one mark was blank), one on either side of the dorsal fin (Hohn & Petrie-Hanson, 2013).
125 Among these marked fish, we used a total of 103 zebrafish with approximately equal sex
126 ratios kept in 4 tanks of 24 individuals (12 males, 12 females) for both experiments. At any
127 one time during the experiments, the same 96 fish were used, but to compensate for death,
128 illness or experimenter error, seven fish were replaced by seven new fish over the three-
129 month study. Due to incomplete data for zebrafish size (described below) the across
130 conditions and within conditions analyses included 93 and 94 zebrafish, respectively. The
131 Garvan Animal Ethics Committee approved all procedures described above and experiments
132 described below (ARA 18_18). Further, Garvan veterinarians oversaw fish welfare associated
133 with aversive learning prior to our pilot tests.

134

135 **Experimental Design**

136 *Aversive Learning Assay*

137 We used an avoidance conditioning method to quantify aversive learning in a simple,
138 automated assay (Brock et al., 2017; Fontana et al., 2019). We ran all assays using four
139 Zantiks AD units (Zantiks Ltd., Cambridge, UK; see <https://osf.io/t95v3/> for further details).
140 The units employed infrared tracking using an integrated computer to record fish movement
141 and collect data. In the assay, a visual cue (colour or pattern) was associated with a negative
142 stimulus (brief mild electric shock; 7V DC 80ms), which motivated fish to avoid the
143 associated visual cue. We then measured the extent of avoidance (i.e., time spent away from
144 the cue associated with an electric shock) compared to the baseline preference to quantify
145 aversive learning (learning response). We based our initial assay parameters (e.g., the
146 acclimation period, voltage, etc) on previous research (Brock et al., 2017), and subsequently
147 modified the parameters based on the outcomes of pilot tests.

148

149 Before each assay we individually placed fish into one of four lanes within rectangular tanks
150 (see Fig. 1A). For the assay, we exposed the fish to four stages; (i) Habituation: we
151 habituated the fish to isolation in a novel environment over a 30-minute acclimation period
152 (Fig. 1B); (ii) Baseline: the tank was visually split into two even zones via the colour
153 displaying screen at the bottom of the tank (Fig. 1C). One of these two colours would later
154 become conditioned with the mild electric shock (CS+), the other colour remained
155 unconditioned (CS-). Here, the position of the colours (left or right) automatically switched
156 every five minutes for a 30-minute period, and we recorded zebrafish preference for the CS+
157 to obtain a baseline preference before conditioning; (iii) Conditioning: first, the CS+ (visual
158 cue associated with shock) was displayed across the entire screen for 1.5 seconds then
159 immediately afterwards paired with the US (mild electric shock) to condition the fish to an
160 aversive experience. Second, the CS- (visual cue not associated with shock) covered the
161 screen for 8.5 seconds (Fig. 1D). This phase was repeated nine times, sufficient for fish
162 learning to avoid the CS+; and (iv) Probe: akin to the baseline period, the tank was split into
163 two even zones (left or right) depicted by different visual cues. We tracked fish movement
164 and recorded fish preference for the visual cue associated with the shock (CS+) over 5
165 minutes. During this time, the visual cues switched every minute (see Fig. 1E). We used only
166 2 minutes out of the 5-minute probe time since we determined a clear decrease in learning
167 response in our observations. This probe length is similar to other studies: Brock et al. (2017)
168 used a 2-minute probe, and Fontana et al. (2019) used a 1 minute probe. Probe CS+
169 preference was used in comparison to baseline CS+ preferences to quantify learning.

170

171 *Experimental Conditions*

172 We used a range of colour conditions to test aversive learning. Each condition was comprised
173 of two visual cues, one aversive and one control (CS+ paired with CS-) (Fig. 1F). We
174 selected different colour combinations to use as visual cues for the zebrafish, which had
175 either been worked in pre-existing assays or were reported to evoke a clear colour preference
176 (Brock et al., 2017; Roy et al., 2019). As a result, we chose seven colours (green, blue, grey,
177 orange, magenta, red, yellow) and 1 pattern (check; hereafter, this pattern is also referred to
178 as a 'colour' with the others). We used four visual cue combinations ('Check/Grey',
179 'Green/Blue', 'Red/Yellow', 'Magenta/Orange') and their reverse ('Grey/Check',
180 'Blue/Green', 'Yellow/Red', 'Orange/Magenta') for a total of eight conditions. For example,
181 the 'Check/Grey' condition used check pattern as the CS+ (cue associated with shock) and
182 grey colour as the CS- (control cue); the 'Grey/Check' condition used grey colour as the CS+
183 and check pattern as the CS-, and so on.

184

185 Prior to the experiment, we assigned fish into quartets (four fish that underwent trials within
186 the same unit/assay tank simultaneously) that systematically rotated between trials. The
187 balanced design accounted for three potential confounding variables: the time of day (quartet
188 rotated), Zantiks unit (quartet rotated), and lane position (individual within quartet rotated).
189 We estimated repeatability in two different situations (across conditions and within a single
190 condition). Across conditions, we ensured fish experienced trials from all four colour pairs
191 before subjecting them to their exact reverse four conditions (with trials conducted over four
192 weeks in June and July 2019). We included this form of reverse learning to negate memory of
193 the CS+ colour between trials, which may impact both baseline and probe colour preference.
194 Within conditions, each zebrafish underwent trials in the 'Blue/Green' and 'Green/Blue'
195 conditions a further two times (over two weeks in September 2019).

196

197 *Fish Size Measuring*

198 We took photos of each fish approximately one week after across-conditions trials and
199 another set of photos approximately one week after within-conditions trials. We captured top
200 down photos of live fish and measured fish in ImageJ (Schindelin et al., 2015). We used fish
201 length (standard length) and width (at widest part of body) to calculate the ellipsoid size of
202 the fish by using $\pi \left(\left(\frac{\text{fish length}}{2} \right) \times \left(\frac{\text{fish width}}{2} \right) \right)$. This controlled for a potential size effect
203 resulting from loss of penetrance and effectiveness of the mild electric shock due to larger
204 body size.

205

206 **Data Processing and Analysis**

207 All data processing and analyses were conducted in the *R* computing environment (version
208 4.0.2; R Core Team, 2019). Linear mixed models were run using the *lme4* package (version
209 1.1.21; Bates et al., 2014) in conjunction with the *lmerTest* package (version 3.1.2;
210 Kuznetsova, Brockhoff, & Christensen, 2017) that provides Satterthwaite's degrees of
211 freedom correction. We obtained repeatability values via the *rptR* package (version 0.9.22;
212 Stoffel et al., 2017) that uses the *lme4* package to run mixed models. Based on visual
213 assessments of residual distributions, assumptions of normality and constant variance were
214 not clearly violated. The Zantiks units recorded time spent in each CS zone, total distance
215 travelled and how often fish changed zones. All code, and the raw and processed data, are
216 available at: <https://osf.io/t95v3/>. We deemed our results statistically significant at the alpha
217 = 0.05 level (or when 95% confidence intervals did not overlap zero).

218

219 *Quantifying Aversive Learning*

220 We determined learning by the difference in time that fish spent in the CS+ before and after
221 the aversive experience. To analyse learning across all the sessions included in this study, we
222 used the time difference (*'difference' = time spent in the CS+ during baseline – time spent in*
223 *the CS+ during probe*) as the response variable in a linear mixed-effects model (LMM) via
224 the *lmer* function in the *lme4* package. We fitted individual 'fish ID' (96 levels) and
225 'experimental condition ID' (8 levels, see Fig. 1F) as random effects in the model. Also, we
226 included the following fixed effects: (1) 'sex' (female or male) to investigate sex differences
227 in learning, (2) 'day' since first trial, to account for time effects of sequential days on
228 learning or learning via repeated trials (e.g., 1 being the first day and 8 being the 7th day from
229 the first), (3) 'fish size' to control for fish's response to conditioning which might be size
230 dependent due to potential differences in body penetrance of a mild shock, (4) 'learning'
231 (initial and reverse) to find if learning was affected when the CS \pm of a condition were
232 switched in successive trials. Note that we z-transformed the fixed effects 'day' and 'fish
233 size' to make the intercept meaningful and slope estimates comparable (Schielzeth, 2010).

234

235 *Quantifying the Repeatability of Aversive Learning*

236 We obtained enhanced agreement repeatability (hereafter referred to as repeatability)
237 estimates by incorporating statistically significant fixed effects from the model and retaining
238 their variance in the denominator (Nakagawa & Schielzeth, 2010). We only fitted the random
239 effect 'fish ID' and included 'sex' as a fixed effect. The R package *rptR* computes
240 repeatability values using the within and between individual variance in linear mixed models
241 fitted with restricted maximum likelihoods (Nakagawa & Schielzeth, 2010). Using *rptR*, we
242 obtained standard errors and 95% confidence intervals (CIs) through parametric
243 bootstrapping, with each model set to 10,000 bootstrap samples. Following Bell (2009) and

244 Wolak (2012), we categorised our repeatability results into low (<0.2), moderate (>0.2 -
245 <0.4) and high (>0.4).

246

247 *Colour Preference and Repeatability*

248 An underlying assumption of our aversive learning assay was that zebrafish can discriminate
249 between different colours. Therefore, from the baseline period (prior to aversive
250 conditioning), we quantified underlying colour preferences (tendency to associate more
251 heavily with one colour in a pair), and the consistency of individual differences in colour
252 preference (i.e., repeatability of colour preference).

253

254 In each condition, preference for one colour was only compared to the other paired colour
255 (e.g., preference for red is only relative to preference for yellow; see Fig. 1F). Given we
256 examined relative colour preference, preferences for either colour in a condition were the
257 inverses of each other. Hence, to be able to determine colour preference for each colour, we
258 grouped conditions of matching colours into four groups for analysis (e.g., Group 1,
259 'Red/Yellow' & 'Yellow/Red'; Group 2, 'Green/Blue' & 'Blue/Green'; Group 3,
260 'Check/Grey' & 'Grey/Check'; Group 4, 'Orange/Magenta' & 'Magenta/Orange').

261

262 To analyse relative colour preference, we ran LMMs for each group of colours using across
263 conditions data. We used baseline colour preference as the response variable 'baseline' for
264 these models. We fitted the random effect 'fish ID' in the models (Group 1 & 4, 97 levels;
265 Group 2 & 3, 98 levels; levels differ because one fish died prior to completing all conditions).
266 Further, we fitted the following fixed effects: (1) 'day' (days since first trial) to control for
267 potential colour preference change with time, (2) 'sex' (male or female) to account for sex
268 differences and (3) 'learning' (initial and reverse) to see the effect of reverse learning on

269 colour preference. To determine the repeatability of colour preference, we used *rptR* mixed-
270 effects models with the response variable ‘baseline’ to generate repeatability estimates. We
271 did not find any fixed effects to be statistically significant, as such, they were excluded, and
272 the colour preference models were fit with the random effect ‘fish ID’.
273

274 **Results & Discussion**

275 We found negligible repeatability in aversive learning across the eight different conditions /
276 colours ($R = 0.04$, 95% CI [0.001 - 0.097], Fig. 2A), despite individuals being able to
277 discriminate between colours (as measured by moderate to high repeatabilities for colour
278 preferences; Grey: $R = 0.45$, 95% CI [0.276 - 0.607]; Green: $R = 0.45$, 95% CI [0.278 -
279 0.604]; Red: $R = 0.43$, 95% CI [0.250 - 0.584]; Orange: $R = 0.46$; 95% CI [0.283 - 0.605]
280 Fig. 2B). Within conditions, we found very low repeatability in one condition (‘Green/Blue’
281 Fig. 2A; $R = 0.02$, 95% CI [0 - 0.153]), and low repeatability in the other (‘Blue/Green’ Fig.
282 2A; $R = 0.15$, 95% CI [0.023 - 0.278]). Therefore, the substantial variation in aversive
283 learning we observed (as in Fig. 3A) was most likely driven by current (intrinsic or extrinsic)
284 environmental factors, rather than additive genetic variance or canalized developmental
285 differences (cf. Sznajder, Sabelis, & Egas, 2012).

286
287 Zebrafish showed strong relative colour preference in all four conditions (see Fig. 3B). We
288 found preference for grey, green, red and orange, over check, blue, yellow and magenta,
289 respectively. The strongest relative colour preference was found for red and orange,
290 providing further evidence that zebrafish prefer colours with longer wavelengths (Roy et al.,
291 2019). We did not find any statistically significant sex difference in colour preference,
292 aversive learning, and their repeatability estimates (see Supplementary Information Table S1-
293 2 and Fig. S1). Of relevance, a previous meta-analysis of repeatability for behavioural traits

294 found males tend to be more repeatable than females (Bell et al., 2009), but reported this
295 result to be inconclusive.

296

297 Our findings of low repeatability for aversive learning are surprising, given low to moderate
298 repeatability of behaviour and cognition reported in two meta-analyses. For general
299 behaviour, Bell et al. (2009) reported an average repeatability of $R = 0.34$. For cognitive
300 performance, Cauchoix et al. (2018) found $R = 0.15-0.28$, mostly based on temporal
301 repeatability estimates from appetitive learning trials. Below we discuss four potential
302 reasons why zebrafish in our experiment showed much less consistent individual differences
303 in aversive learning compared to the previous estimates from Cauchoix et al. (2018) and Bell
304 et al. (2009).

305

306 First, while zebrafish did demonstrate aversive learning, the average effect was small, and in
307 many trials, individuals did not seem to avoid the negative stimulus. On average, individuals
308 spent just 4-6 fewer seconds per minute in the negatively associated colour following
309 conditioning (across conditions: female average = 3.89 seconds per min, $SE = 1.05$, $t_{33} =$
310 3.65 , $P < 0.001$; male average = 5.64 seconds per min, $SE = 0.94$, $t_{22} = 5.21$, $P < 0.001$; Fig.
311 3A). The small effect could be caused by individuals not learning or quickly forgetting. It is
312 also possible that learning performance would be greater at the group level; zebrafish are a
313 shoaling species and learning may have evolved to depend on group dynamics. When tested
314 individually, zebrafish display more variable behaviour and are more prone to stress
315 (Pagnussat et al., 2013). Low repeatability could therefore be caused by zebrafish being
316 largely insensitive or unresponsive to the conditioning when housed individually (i.e., poor
317 aversive learners, a weak assay or anxious fish with impeded movement). However, the fact
318 that there was a population shift in the direction of aversive learning raises the question of

319 why individuals who learnt in one trial did not maintain their performance across trials; if a
320 particular subset of zebrafish had consistently learnt, or failed to learn, then we would have
321 detected higher repeatability. Further, while the behaviour change following aversive
322 conditioning was modest, zebrafish learnt much faster (in 1.5 minutes) compared to previous
323 assays with appetitive training (e.g., over 20 days; Brock et al., 2017). As far as we are
324 aware, no studies have investigated a relationship between the strength of associative learning
325 and the magnitude of repeatability. Furthermore, it should be noted that our experiment only
326 considered visual cues, but in the wild, fish often use chemical cues to detect danger (Brown,
327 2003). Although technically more challenging, aversive learning using different sensory cues
328 other than visual cues should be considered in the future.

329

330 Second, past selection pressures on our study population may have eroded additive genetic
331 variance associated with aversive learning, which was not restored in the intervening
332 generations. In the wild, aversive learning could be under strong selection (e.g., to learn to
333 evade predators), and individuals could be selected to learn from negative experiences as
334 quickly as possible. Indeed, aversive learning could be under stronger selection than
335 appetitive learning, as mortality costs of negative experiences can easily exceed opportunity
336 costs of missing positive experiences. Stronger selective pressures could explain why we
337 found substantially lower repeatability for aversive learning compared with previous results
338 for appetitive learning. In a similar vein, traits more closely associated with fitness (e.g.,
339 aversive learning) tend to not be as heritable (thus, repeatable; cf. Dohm, 2002) than less
340 fitness related traits (e.g., appetitive learning; Merilä & Sheldon, 2000). However, we cannot
341 be sure that the performance of zebrafish in our laboratory assay accurately captures their
342 ability to aversively learn in their natural habitat.

343

344 Third, some of the repeatability values in the meta-analyses by Cauchoix et al. (2018) and
345 Bell et al. (2009) may have been overestimated. An inflated repeatability estimate, also
346 known as ‘pseudo-repeatability’, is the result of within-individual variation being erroneously
347 accredited to differences between individuals (Niemelä & Dingemanse, 2017; Westneat et al.,
348 2011). Pseudo-repeatability occurs when the conditions between measurements are too
349 similar (e.g., environmental conditions are unchanged or intervals between measurements are
350 too short) and might explain why we found higher repeatability when zebrafish were
351 measured repeatedly within a single condition (‘Blue/Green’; $R = 0.15$), compared to across
352 eight separate conditions (although no inflation was seen in ‘Green/Blue’). Indeed, Cauchoix
353 et al. (2018) and Bell et al. (2009) included studies with testing conditions which did not
354 change over the course of a study, similar to our within-condition estimates. Further, most
355 studies in both meta-analyses had relatively short intervals between measurements (most
356 intervals were under a week in Cauchoix et al. (2018), and almost all were under a year in
357 Bell et al. (2009). The short intervals between measurements reported in Bell et al. (2009)
358 were significantly associated with higher repeatability values, consistent with pseudo-
359 repeatability. Relevantly, two recent studies on birdsong reported that associative learning
360 among individuals was not repeatable between years, indicating that estimates obtained over
361 short intervals may not be a true reflection of consistent individual differences defined in
362 animal personality (Soha et al., 2019; Zsebök et al., 2017).

363

364 Fourth, publication bias might have contributed to an inflation of the overall repeatability
365 estimates in the published literature (cf. Parker et al., 2016). The average repeatability of 0.34
366 reporting by Bell et al. (2009) was based on a meta-analysis of published studies. Cauchoix et

367 al. (2018) included many more unpublished datasets (n = 38) compared to published datasets
368 (n = 6); they mentioned that their unpublished datasets produced, overall, a lower
369 repeatability estimate than that of the published studies. This finding is consistent with the
370 pattern that larger effect sizes are more likely to be published. Recent studies are increasingly
371 reporting non-significant and low repeatability (e.g., Reichert et al., 2020; Vernouillet &
372 Kelly, 2020). Therefore, an updated future meta-analysis may reveal a lower overall
373 repeatability estimate in behaviour.

374

375 In conclusion, zebrafish did not show clear consistent between-individual differences in
376 aversive learning. The low repeatability could potentially indicate that strong past selection
377 pressure has almost driven aversive learning to fixation, because of the vital importance of
378 learning to avoid danger. Alternatively, low repeatability may be due to the small effect of
379 fish learning to avoid the stimuli. In addition, published repeatability estimates could be
380 inflated by within-individual variance frequently being measured as between-individual
381 differences (i.e., ‘pseudo-repeatability’), and by publication bias. We contend that these
382 issues can be diminished in future behavioural research by controlling for confounding
383 effects and reporting every estimate of behavioural traits, whether repeatable or not.

384

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517 **Figure Legends**

518 **Figure 1**

519 **Colour conditions and aversive learning assay:** (A) zebrafish are placed in the
520 experimental tanks and (B) acclimated to the novel environment for 30-minutes; (C) in a 30-
521 minute baseline period, initial CS± preference is established; (D) during the conditioning

522 phase, fish are presented the CS+, then immediately subjected to a mild electric shock; and
523 (E) in a 5-minute probe phase, learning is determined by fish spending less time in the CS+
524 when compared to the baseline. (F) Each condition is a combination of two visual cues
525 (zones), one conditioned to a mild electric shock (CS+), the other is not (CS-). Across
526 conditions eight colour conditions and eight sessions (each session is represented by a white
527 box). Within conditions: two colour conditions and four sessions (in addition to two sessions
528 in Experiment 1).

529

530 **Figure 2**

531 **Repeatability of aversive learning and colour preference in zebrafish.** Points and
532 whiskers represent means and 95% confidence intervals via parametric bootstrapping. (A)
533 Zebrafish show somewhat consistent individual differences in aversive learning within the
534 'Blue/Green' pair, but not within the 'Green/Blue' pair or across all colour combinations. (B)
535 Zebrafish show consistent individual differences in colour preferences (variation depicted in
536 Fig. 3B).

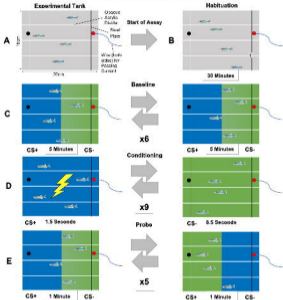
537

538 **Figure 3**

539 **Violin plots for aversive learning and colour preferences.** Smaller coloured points depict
540 individual trials. Larger black points and error bars depict means and standard deviations of
541 observations. (A) The top panel shows means and variation in aversive learning, split by sex
542 (female = purple; male = green) when all the session data is combined. Points above the line
543 at zero depict trials in which zebrafish spent less time in the aversive stimulus colour in the
544 probe period (the colour associated with an electric shock) relative to the baseline period (i.e.,
545 aversive learning). (B) The bottom panel shows the tendency of zebrafish to favour one
546 colour in a pair during the baseline period (i.e., before administration of electric shocks). The

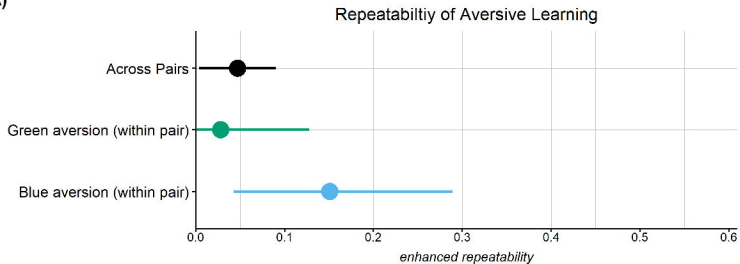
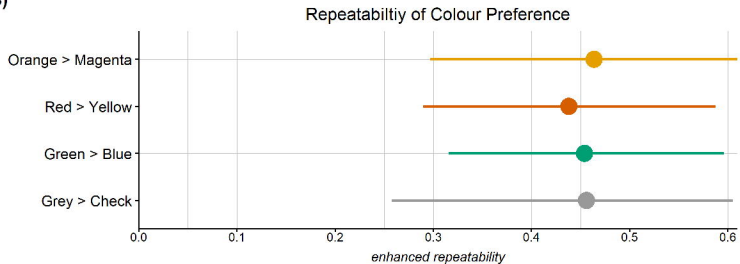
547 dashed horizontal line at 0.5 represents no colour preference (i.e., spending 30 seconds in
548 each colour zone).

Avoidance Conditioning Assay



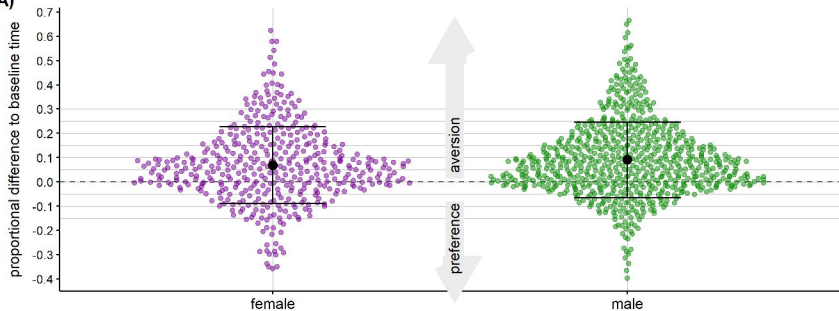
Experimental Conditions



(A)**(B)**

Aversive Learning

(A)



Baseline Colour Preferences

(B)

