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How stable are personalities? A multivariate view of behavioural variation over long and short timescales in the sheepshead swordtail, Xiphophorus birchmanni

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1	How stable are personalities? A multivariate view of behavioural variation over
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4	
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14 Abstract

15 Many studies have revealed repeatable (among-individual) variance in behavioural traits consistent 16 with variation in animal personality; however, these studies are often conducted using data collected over single sampling periods, most commonly with short time intervals between 17 18 observations. Consequently, it is not clear whether population-level patterns of behavioural variation are stable across longer timescales and/or multiple sampling periods, or whether 19 individuals maintain consistent ranking of behaviours (and/or personality) over their lifetimes. Here 20 21 we address these questions in a captive bred population of a tropical freshwater poeciliid fish, 22 Xiphophorus birchmanni. Using a multivariate approach, we estimate the among-individual variance-covariance matrix (I), for a set of behavioural traits repeatedly assayed in two different 23 experimental contexts (open field trials, emergence and exploration trials) over long- (56 days 24 25 between observations) and short-term (four day observation interval) time periods. In both longand short-term data sets we find that traits are repeatable and the correlation structure of I is 26 27 consistent with a latent axis of variation in boldness. While there are some qualitative differences in 28 the way individual traits contribute to boldness, and a tendency towards higher repeatabilities in the short term study, overall we find that population-level patterns of among-individual behavioural 29 (co)variance to be broadly similar over both time frames. At the individual level we find evidence 30 that short-term studies can be informative for an individual's behavioural phenotype over longer 31 32 (e.g. lifetime) periods. However statistical support is somewhat mixed and, at least for some observed behaviours, relative rankings of individual performance change significantly between data 33 34 sets.

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39 Introduction

40 It is now apparent that, within animal populations, individuals often exhibit differences in behaviour 41 that are repeatable across time and context. This repeatable variation is taken as evidence for animal temperament (e.g. Boissy 1995; Réale et al. 2007), behavioural syndromes (Sih et al. 2004), 42 43 coping styles (Koolhaas et al. 1999), or personality, the latter term reflecting parallels with research in human psychology (Budaev 1997b; Gosling 2001). A number of axes of among-individual 44 behavioural variation condensed into "personality traits" have been described, including boldness-45 46 shyness, exploration-avoidance and general activity (Réale et al. 2007). Understanding the evolution 47 of personality has become a major field of study in behavioural ecology (Dall et al. 2004; Stamps and Groothuis 2010). There is now growing evidence that traits relating to personality contribute to 48 fitness variation and therefore may be both adaptive and generally under selection (Smith and 49 50 Blumstein 2008). However, if natural selection occurs through variation in lifetime fitness, then an important question arises: just how stable are personalities over individual lifetimes? Here we 51 52 address this question in a captive population of fish. We do this using a novel multivariate approach that characterises personality variation as a latent character underpinning among-individual 53 (co)variation in a suite of observed behaviours. 54

While there remains considerable disagreement over how best to define individual personality traits 55 56 (Réale et al. 2007; Toms et al. 2010; Carter et al. 2013; see below) there is broad consensus that among-individual behavioural variance is the statistical signature of animal personality. Typically this 57 is quantified as the (among-individual) repeatability, defined as the proportion of observed variance 58 59 explained by individual identity, of one or more observed behavioural traits. Thus partitioning of 60 observed variance into among- and within-individual components (the latter arising from individual plasticity and/or measurement errors) from repeat observations on individuals is crucial to empirical 61 studies of personality (Dingemanse et al. 2012b; Brommer 2013; Araya-Ajoy and Dingemanse 2014). 62 63 In a meta-analysis, Bell et al (2009) concluded that on average, estimates of repeatability for

observed behavioural traits decreased as the interval between sampling events increased. 64 Consequently, it may be dangerous to assume that short-term studies reflect behavioural (and by 65 implication, personality) differences that are stable over the lifetime of individuals. This is 66 67 potentially important since short-term repeatability estimates predominate in the literature, although the number of studies conducted over timeframes that may be considered more 68 69 representative of natural life-spans is growing (for more recent examples, see Ronning et al. 2005; 70 Bushuev et al. 2010; Chervet et al. 2011; David et al. 2012; Kanda et al. 2012). However, few studies 71 have collected repeated observations over two distinct time periods from the same individual (but see for e.g. Carere et al. 2005) that would allow the "repeatability of repeatability" to be assessed. 72 73 Here we do this, but also extend our analysis to the multivariate case to ask whether patterns of 74 among-individual behavioural (co)variation reflect an underlying personality trait that is stable 75 across distinct long- and short-term sampling periods.

In what follows we investigate the temporal stability of multiple behavioural traits in the freshwater 76 77 poeciliid fish, Xiphophorus birchmanni to answer two complementary questions. Firstly, at the level 78 of the population, how stable are the patterns of among-individual trait (co)variance generated by 79 underlying personality? Secondly, at the level of the individual, do short term studies reveal 80 behavioural tendencies that are stable across lifetimes? To answer these questions we characterise behavioural variation along what we loosely consider to be an axis of shyness-boldness. Boldness is 81 82 the most commonly studied axis of personality in fish (Toms et al. 2010), and positively correlates 83 with fitness-related traits including reproductive success, parental provisioning, growth, aggression, social dominance, dispersal and proactive responses to stressors such as predation risk (Dingemanse 84 et al. 2004; Brown et al. 2005; Bell and Sih 2007; Cote et al. 2010; Rudin and Briffa 2011; Ariyomo 85 86 and Watt 2012; Mutzel et al. 2013). There remains, however, a lack of consensus on how best to define boldness and how it should be assayed (Toms et al. 2010). This raises obvious potential for 87 misclassification of personality traits (Carter et al. 2013), and/or disagreement over appropriate 88 89 experimental design (Toms et al. 2010).

The present goal is to investigate stability of a personality trait without adding further to existing 90 91 debate over issues of definition. Consequently we do not attempt to define boldness or the best way to measure it *a priori*; rather, we follow the view of others that personality traits should be 92 considered as latent variables that can best be uncovered by observing several measurable, 93 correlated and potentially overlapping behaviours across contexts (Dochtermann and Jenkins 2007; 94 95 Dingemanse et al. 2010; Dochtermann and Roff 2010). We therefore make a distinction throughout between "behavioural traits" that are observed directly, and "personality (traits)", inferred from 96 97 among-individual (co)variance in observed behaviour(s). This exploratory approach, which follows Huntingford (1976) and others (Budaev 1997b; Moretz 2003) is becoming more mainstream and 98 99 allows the avoidance of difficulties that can arise if a single behaviour is chosen a priori to assay 100 boldness. For example, a fish that swims a long distance in one behavioural trial may be classified as 101 willing to explore and therefore as "bold"; however, this behaviour could also plausibly be indicative of anxiety, with the animal's "exploration" being driven by a search for refuge. 102

103 Currently the most common experimental paradigm used to measure boldness is that of the open 104 field trial (OFT), in which an animal is placed in an open arena and its behaviour is monitored for a 105 predetermined observation period. Initially developed for rodent studies (Hall 1934; Walsh and Cummins 1976), OFTs have long been applied to fish models (Warren and Callaghan 1975; Budaev 106 1997b). Considered the most reliable way to assay boldness by some authors (Burns 2008), others 107 108 have argued that OFTs risk conflating boldness with other axes of variation that are distinct (if 109 sometimes correlated) personality traits in their own right (e.g. exploration-avoidance, overall activity, Réale et al. 2007). If so, then simple modifications to OFTs such as providing a refuge from 110 which an animal can choose to emerge and explore (emergence and exploration trial, EET) may be 111 useful (Dingemanse et al. 2007). 112

In what follows we use both types of behavioural trial mentioned above (OFT and EET) to observe
how fish behave in these contexts and to characterise the repeatable component of multivariate

behaviour. We then assess the extent to which one or more major axes of variance adequately 115 depict observed variation. In other words, we aim to describe the behavioural trait variation first, 116 and then consider the extent to which its repeatable component fits within the paradigm of a major 117 axis of personality, i.e. the boldness-shyness axis (Dingemanse et al. 2010; Dochtermann and Roff 118 2010). We then go on to address three specific questions regarding the temporal stability of 119 120 personality. Firstly we ask whether repeatabilities estimated from repeated measures of individual behaviours over a short time period give a misleading view of the importance of among-individual 121 122 variance over longer time periods. Secondly, by extending our analysis to the multivariate case we ask whether the structure of the between-trait among-individual covariance matrix, denoted I, 123 124 following Wilson et al. (2013), is similar when estimated from short- and long-term data; i.e. do 125 repeated empirical analyses of a single population actually reveal the same major axes of amongindividual variation? If so, then a final question concerns the extent that individuals retain the same 126 relative ranking for repeatable behaviours, and hence personality, over their lifetimes. 127

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129 Methods

130 Study species and husbandry

One hundred wild adult Xiphophorus birchmanni were caught in the Arroyo Coacuilco near the town 131 of Coacuilco, municipality of San Felipe Orizatlán, Hidalgo, Mexico, (elevation 314 m lat/long 21.099 132 -98.587), and imported to the UK in February 2010. Between August 2010 and May 2011 we 133 collected an offspring generation (n = 384) from 13 males and 27 females (mean (SE) brood size of 134 8.86 (0.541)). Gravid females were isolated and, following birth, broods were immediately netted 135 and moved to one half of a partitioned 30 L tank; broods of more than six offspring were split with 136 each half of the family placed in different tanks. Fry were fed twice daily on a mix comprising equal 137 138 quantities of crushed ZM spirulina and brine shrimp flake and laboratory prepared brine shrimp

nauplii. At an average of 17 weeks (range 12 to 27) juveniles were tagged with a single elastomer 139 140 injection for individual identification purposes and transferred to mixed-family rearing groups of n = 8. Note it is not possible to determine sex at this age in this species and therefore the sex ratio was 141 not controlled. Eight rearing groups were then kept within each of six sequentially set-up stacks of 142 tanks, each stack sharing a common water supply and recirculating filtration system. As part of a 143 144 parallel study of density effects on growth, rearing groups were initially housed under two different density regimes as follows. Within each stack, four groups were placed in 30 litre (37 x 37 x 22 cm) 145 146 glass tanks (low density treatment) with the remaining four groups in 15 L half tanks (high density treatment). Half tanks were created by placing a black net covered Perspex-framed partition down 147 148 the centre of a full – size tank. Thus, establishing a stack required 64 fish (i.e., 8 x 8) to be available 149 for tagging simultaneously and this accounts for the variation in tagging age within stacks. Fish were fed twice daily with a standardised ration of flake food as above (morning) and a mix of previously 150 151 frozen blood worm and daphnia (afternoon). On the days when behavioural data was to be collected, the morning feed was omitted in an attempt to encourage exploration tendencies. 152 Temperature was maintained between 22 - 24°C and a 12:12 hr light:dark cycle imposed. After 153 being housed in this manner for 28 weeks, density was swapped for half of the tanks, thus creating 154 four treatment effects with the total number of fish divided approximately equally between them as 155 156 follows: Low/Low (n = 93), Low/High (n = 95), High/High (n = 87), High/Low (n = 93). Observations from individuals failing to reach sexual maturity by the end of the long-term study (50 weeks), were 157 158 excluded from the analysis and the above breakdown (n = 11).

159 Behavioural data collection

The trials were performed over two experimental study periods, denoted long-term (LT) and shortterm (ST). All available fish contribute to the long-term data set (n = 373) while a random subset of 32 fish from each of the four density treatments (Low/Low n = 13, Low/High n = 4, High/High n = 9, High/Low n = 6) was used for the short-term study (Table 1). Trials were of two types, open field

(OFT) and emergence and exploration (EET) with multiple specific behavioural traits assayed in each 164 165 trial type (Table 2). Overall, the long-term trials took 13 months to complete (May 2011 – May 2012), with data collected over an actual 30 week period for each fish. Each individual was subject 166 to an OFT followed by an EET seven days later, a process that was repeated three times at 56 day 167 intervals, thus yielding four OFT and four EET trials per fish. The short-term data set was collected in 168 169 February 2013, with 32 individual fish subjected to alternating OFT and EET at 48 hour intervals (i.e. 2 days between trials, 4 days between repeated trials of the same type) with each animal 170 171 undergoing five trials of each type. For those 32 individuals used in both study periods data was therefore collected over a timeframe with a mean (SE) of 531.4 (6.38) days. By comparison the 172 mean (SE) longevity of individuals with known birth and death dates under our laboratory conditions 173 is 450.3 (8.10). 174

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176 Experimental procedures

177 Open Field Trial (OFT)

An empty 45 x 25 x 25 cm tank was filled to a depth of 8 cm with room temperature water (22° C).

179 The tank was lit from below and visually screened by a cardboard casing to prevent external

180 laboratory disturbance. Fish were caught individually from their home tank with a dip net, quickly

181 examined for identification tags and immediately placed into the centre of the OFT tank. Following a

182 30 second acclimation period, behaviour was filmed for 300 seconds using a Sunkwang C160 video

camera fitted with a 5 – 50 mm manual focus lens suspended above the tank. Data were then

184 extracted from the video using the tracking software Viewer II

185 (http://www.biobserve.com/products/viewer/index.html), which was set up to divide the tank basal

area into two approximately equal halves (middle and perimeter zones) (Fig. 1a). Water was

187 changed between individual trials to prevent chemical cues affecting behaviour.

188 Emergence and Exploration Trial (EET)

189 A 45 x 25 x25 cm tank was physically divided into three sections with opaque Perspex, providing a 190 right-hand, centre and left-hand chamber. A small (5 cm) opening was cut in each divider, starting two cm from the tank edge. The openings were positioned at opposite sides of the tank. The 191 192 chamber on the right hand side was designated as the refuge, and equipped with a plastic plant and several small stones. A rising trapdoor was rigged to a pulley above the tank, and positioned inside 193 the refuge and covering the exit into zone 1 (Fig. 1b). Tanks were filled, emptied, lit and screened as 194 195 above. Fish were individually caught and examined as before, and placed directly into the centre of the refuge where they were allowed 30 seconds to acclimate before the trapdoor was lifted. Filming 196 then commenced for 300 seconds (as above), but only behaviour outside the refuge (i.e. in zones 1 197 and 2) was tracked and extracted for analysis. 198

199 Behavioural traits

200 The behavioural traits recorded in this study were selected as those likely to reflect variation along a 201 bold-shy type personality axis. For the OFT, we predicted that fish tending toward boldness would 202 actively explore the novel environment of the OFT by leaving the tank sides and spending more time 203 in the central zone than shy fish. OFT behaviour was therefore quantified by four traits; Track 204 Length (TL), Activity (Act), Area Covered (AC) and Time in Middle of the tank (TIM), which we 205 predicted would be positively correlated with one another. In the EET, we expected bold fish to locate the doorway in the refuge and leave through it. We recorded two traits from the EET: 206 whether or not the individual emerged from the relative safety of the refuge (Emergence) and 207 Latency in seconds to do so. We predicted positive within-individual correlations between 208 Emergence from the refuge and the OFT traits, with negative correlations between Latency to 209 Emerge and all other traits. Note that the EET tank was set up with the area outside the refuge 210 211 further divided into two zones (1 and 2; Fig 1b). In the EET, we had initially planned to use "latency

to enter zone 2" (distal to the refuge) as an additional trait in our analyses; however, in practice this
became a redundant trait due to a low frequency of fish entering this area.

214 Statistical analyses

215 All data were modelled using restricted maximum likelihood mixed effects models implemented in ASReml V3 (Gilmour et al. 2009). Prior to analysis, data for the OFT trait Time in Middle were square 216 root transformed to reduce positive skew. Visual inspection of residuals suggested that the 217 assumption of residual normality was reasonable for the other traits recorded in OFT. All traits were 218 219 rescaled to standard deviation units prior to analysis to prevent trait scale effects from influencing the structure of I (defined and estimated as described below). Given that a large proportion of fish 220 did not emerge from the refuge (see results) the Latency to Emerge data were heavily censored and 221 we elected to use only the binary variable of Emergence in subsequent analyses. Emergence was 222 included in full multivariate models using REML under an assumption of (multivariate) residual 223 224 normality. Statistical inferences on this trait should therefore be treated with obvious caution.

225 While statistical approaches exist that allow non-Gaussian trait distributions to be used (e.g. MCMC 226 Bayesian approaches implemented in the R package MCMCgImm (Hadfield, 2010)) they do not currently allow the error structures appropriate to our multivariate models (i.e. no definable or 227 228 estimable residual covariance between OFT and EET traits – see below) and thus could not be used 229 here. However, we checked the validity of REML-based conclusions regarding Emergence by fitting additional univariate and bivariate models using MCMCgImm. Specifically we fitted a univariate 230 model of Emergence to estimate the repeatability of this trait and bivariate models of Emergence 231 with all other OFT traits to estimate the covariance structure between these traits. All models in 232 MCMCgImm modelled Emergence as a categorical trait with the residual variance fixed at 1 and all 233 OFT traits as Gaussian. All MCMCglmm models were run for a total of 1050000 iterations with a 234 235 burnin of 50000 iterations and a thinning interval of 1000 iterations. The repeatability of Emergence from MCMCglmm models was defined as the intraclass correlation, calculated as V₁ / (V₁ + V_R + $\pi^{2/3}$), 236

where V_1 is the among-individual variance and V_R is the residual variance that in this case is fixed to 1 (Hadfield 2010).

239 To test the hypothesis that among-individual variance for behavioural traits is both present and repeatable in our fish species, we first combined data from both collection periods and fitted a 240 241 multivariate model of our observed behavioural traits. For each trait we included fixed effects of the mean, sex (a two level factor determined from external morphology at maturation), home stack (a 242 six level factor accounting for differences between sets of fish sharing water supplies), trial number, 243 244 density treatment, and day order. Trial number is the cumulative number of trials experienced by an individual (fitted as a linear effect). Density treatment is a four level factor describing density 245 conditions experienced in the rearing stacks. Day order was modelled as a linear effect of the 246 number of preceding trials conducted on any day and was used as a proxy for time of day. This was 247 included to control for potential diurnal rhythms in fish behaviour. We also fitted an interaction 248 term of trial number * density treatment, in case any systematic changes in observed trait means 249 250 across trials (due to e.g., age effects, habituation etc.) are themselves treatment dependent. Wald F-tests were used to test the significance of fixed effects in the models. 251

252 By including individual identity as a random effect, we then partitioned multivariate phenotypic (co)variance not explained by the fixed effects into an among-individual and a within-individual 253 254 (residual) component. The former is estimated as the variance-covariance matrix I, which contains estimates of the among-individual variance (V_1) component for each trait on the diagonal and 255 estimates of the corresponding covariance between trait pairs (COV_1) off the diagonal. The within-256 257 individual component is similarly estimated as a residual variance-covariance matrix (R). We make the standard assumptions that residual errors are normally distributed and uncorrelated across 258 observations, and that (co)variance parameters in I and R are homogeneous across levels of the 259 fixed effects (i.e. density treatments, trial number, stack etc). Although the two experiment-specific 260 261 sets of traits are not observed in the same trials, we grouped the data by trial period, (e.g. OFT1 with EET1). Thus, we modelled a residual covariance term between OFT and EET traits observed within each trial period. Repeatability (R₁) was then estimated for each trait as the among-individual variance (V₁) divided by total phenotypic variance (V_p) (where V_p is the phenotypic variance conditional on the fixed effects; i.e. V_p = V₁ + V_R). Between each pair of traits (1, 2) the amongindividual covariance (COV₁) was rescaled to give the corresponding correlation r₁ (where r_{1(1,2)} = $COV_{1(1,2)} / v(V_{11}^* V_{12})$).

To test the statistical significance of among-individual behavioural variation we compared the 268 269 likelihood of our full multivariate model to two further models. In the first of these, we fitted I as a 270 diagonal matrix such that the model allows among-individual variance V_1 for each trait, but assumes Cov_1 is zero between all trait pairs. In the second, a null model, we removed the random effect of 271 individual identity completely. Comparison of the diagonal model with the null model using 272 273 likelihood ratio tests (LRT) allows a global test of the significance of among individual behavioural variance (Wilson et al. 2010). Comparison of the full model with the diagonal model, again by LRT, 274 275 allows a statistical test of whether I contains significant between-trait covariance structure (Wilson et al. 2013). LRT were performed by estimating χ^2_{nDF} as twice the difference in model log likelihoods, 276 with the number of degrees of freedom (n) equal to the number of additional parameters to be 277 278 estimated in the more complex model.

279 The above analyses were then repeated using long- and short-term data subsets to estimate the corresponding matrices I_{LT} and I_{ST} and associated parameters. Note that, following the conclusion of 280 the LT, the density treatments were no longer applied and the 32 fish used in the ST were housed 281 282 together in the same stack. Therefore, the fixed effect *stack* was redundant and omitted from the models for the short-term subset analyses. To further investigate the structure of I, I_{LT} and I_{ST} , we 283 subjected each matrix to eigenvector (EV) decomposition. This allowed us to examine: a) how much 284 variance is captured by the first axis (EV1) of multivariate behaviour in each case, b) whether factor 285 loadings of individual traits onto EV1 are consistent with an interpretation of "boldness-shyness" 286

and c) whether EV1 is similar in I_{LT} and I_{ST} . To provide a quantitative measure of how similar the multivariate behavioural axes emerging from the long- and short-term data sets were, we calculated the angle (θ) between the first eigenvectors of I_{LT} and I_{ST} . An angle of $\theta = 0^{\circ}$ equates to the vectors being perfectly aligned, meaning that EV1, i.e. the axes of multivariate behavioural variation in I_{LT} and I_{ST} are identical. Conversely, an angle of $\theta = 90^{\circ}$ would indicate the vectors are orthogonal (and thus maximally differentiated) to each other across the two different time periods (i.e. the major axis of behavioural variation across the two studies are independent).

294 Uncertainty around the factor loadings for individual traits on EV1 (for I matrix) and around θ was estimated using a parametric bootstrap approach (similar to that outlined in the appendix of 295 (Morrissey et al. 2012)). We simulated 5000 replicate draws of I, I_{LT} and I_{ST} from multivariate normal 296 distributions using the maximum likelihood estimates of these matrices as the means, and the 297 298 variance-covariance matrices of their elements to define the variances. In each case the 5000 simulated matrices were subject to Eigen decomposition. Uncertainty around the point estimates of 299 300 trait-specific factor loadings was then described using the 95% highest probability density interval the simulated values of these loadings (for I, I_{LT} and I_{ST} respectively). Note that these intervals 301 should be viewed as approximate as they are vulnerable to departures from multivariate normal 302 303 assumptions. By comparing 5000 pairs of simulated LT and ST matrices we similarly estimated the uncertainty around our point estimate of θ . Note however that since θ cannot be less than zero, we 304 305 also generated a null distribution for the estimator in the absence of any difference between (true) I 306 matrices. This was done by comparing the leading eigenvector of each of the 500 replicate draws of I_{LT} (simulated as described above), to the leading eigenvector of a second matrix, simulated with the 307 same mean (i.e. the REML point estimate of I_{LT}) but a variance equal to the estimated variance-308 309 covariance matrix from the short-term study. Thus the null distribution represents θ estimates given 310 that i) the angle is zero since true I matrices are identical (and equal to the REML estimate of I_{LT} , but ii) the second (short-term) matrix (and so its leading eigenvector) is estimated with greater 311 312 uncertainty due to the lower sample size.

Finally, we compared V₁ estimates in LT and ST data subsets, and tested the among-individual, across 313 data subset correlations $(r_{I(LT,ST)})$. For each behavioural trait (x) we used a likelihood ratio test to 314 compare a bivariate model of x_{LT} and x_{ST} where V_1 is constrained to be equal, to a model where it is 315 free to vary. This tests the hypothesis that among individual variance differs across data sets. (Note 316 that since traits are analysed in observed standard deviation units V₁ can also be interpreted as the 317 318 repeatability estimate unconditional on fixed effects). We then expanded this model to estimate the among-individual, across data subset correlation $(r_{I(LT,ST)})$ and tested this against null hypotheses of 319 320 both $r_1 = 0$ and $r_1 = +1$. Estimation of this correlation is possible since the 32 fish used in the shortterm study were a subset of the long-term study. If $r_1 = +1$, then this indicates that the ranking of 321 322 phenotypic merits (i.e. each individual's repeatable component of the observed trait) is the same across data sets. However, if $r_1 = 0$, then an individual phenotypic merit in the long-term study is 323 uncorrelated with the repeatable component of that same behaviour observed over a short time 324 period in later life. 325

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327 Results

In total, 1235 sets of behavioural observations were conducted from a possible 1492, the difference being due to mortality of some fish over the study period. Summary data for all behavioural traits are presented in supplemental materials, Fig. S1. In EET, the number of fish emerging from the refuge within the observation period was lower than anticipated based on pilot data (LT = 526/2448, ST = 100/318), resulting in severe censoring of Latency to Emerge data. We therefore elected to use only the binary Emergence trait from this trial type in our analyses.

334 Analysis of full data set

There was significant among-individual variance in multivariate behaviour (diagonal model versus null model, $\chi^2_5 = 125.6$, P<0.001), as well as among-individual covariance among traits (diagonal model versus full model, $\chi^2_{10} = 101.8$, P<0.001). Estimates of individual repeatability (R₁ (±SE)) were low to moderate, ranging across traits from 0.055 (±0.024) for Emergence (on the observed scale, estimated by REML) to 0. 192 (±0.029) for Time in Middle (Table 3). Based on univariate models, V₁ was statistically significant at P<0.05 for all traits (Supplemental Table S2). The estimated fixed effects are not directly relevant to present objectives; however they are presented in full in the supplemental materials (Supplemental Table S3).

Between traits, the signs of all among-individual correlations (r₁) were positive, consistent with our *a priori* expectations (Table 3). The OFT traits Track Length, Activity and Area Covered were all strongly correlated (and nominally significant based on $|r_1| >$ two standard errors); however while Time in Middle was strongly correlated with Area Covered ($r_1 = 0.653 \pm 0.075$, Table 3), it was only weakly associated with the other OFT traits. The EET trait Emergence was positively correlated with each OFT trait (r_1 estimates ranging from 0.304 with Track Length to 0.577 with Activity, Table 3).

Eigen analysis of I, estimated from the full data set revealed that the first two vectors explained 64 % 349 350 (eigenvector 1, EV1) and 26 % (eigenvector 2, EV2) of the repeatable among-individual variation 351 respectively (Fig. 2). The trait loadings on the dominant vector EV1 are consistent with an 352 interpretation of this axis of variation as boldness (or arguably exploration and/or general activity; see discussion). Thus individuals that tended to emerge repeatedly in the EET, swim longer 353 354 distances, are more active explore more area, and spend more time in the middle of the OFT tank. By comparison, EV2 trait loadings show this axis to be dominated by time spent in the middle of the 355 tank. Track Length and Activity load on this vector to a lesser extent and with an opposing sign to 356 357 Time in Middle, while the other traits show limited contributions to EV2 (Fig. 2b).

As noted earlier, our REML analysis makes an assumption of (multivariate) residual normality that is
 violated by inclusion of the binary trait Emergence. Univariate analysis of Emergence using
 MCMCglmm, calculated following equation 15 of (Nakagawa and Schielzeth 2010), yielded a slightly
 higher estimate of repeatability (on the liability scale) with a posterior mode of R = 0.090, 95% HPD

interval 0.024 – 0.177, Table S1). While noting that interval will never span zero since R is
constrained to lie in positive parameter space, the posterior mode is nonetheless "distinct" from
zero (Supplemental material, Fig. S2). Bivariate models (i.e. the use of one OFT trait plus Emergence
as the phenotypic variates) also confirmed the presence of strong positive among-individual
correlations (r₁) between Emergence and OFT traits. Thus, the MCMCglmm analyses corroborate the
results of the REML analysis for Emergence (Supplemental material, Table S1).

368 Comparison of long- and short-term results.

In both the long- and short-term studies, the presence of repeatable variance was statistically supported (comparisons of null and diagonal model: $LT \chi_5^2 = 77.0$, P < 0.001; $ST \chi_5^2 = 29.7$, P < 0.001) as was the presence of between-trait among-individual covariance structure (comparisons of diagonal and full multivariate model: $LT \chi_{10}^2 = 95.0$, P < 0.001; $ST \chi_{10}^2 = 54.9$, P < 0.001). Univariate models confirmed that V₁ was statistically significant for all OFT traits in both LT and ST, but not for Emergence in ST (Supplementary Tables S2).

The estimate of I_{LT} is very similar to that obtained using all data (as described above), not 375 376 unexpected given that the long-term study contributes the bulk of the total data set. However, comparison of ILT and IST (and derived parameters thereof) indicates some differences in the 377 378 structure of among-individual behavioural variation as estimated from our long- and short-term 379 studies (Table 3). Note that the smaller size of the short-term data set means that the estimates are less precise for this study; this is reflected in the larger standard errors associated with the 380 parameters. Repeatability estimates (R) were higher in the short term study across all traits. 381 However the increased R from ST was particularly striking for Track Length (Table 3, Fig. 3). For this 382 trait, along with Activity and Area covered the null hypothesis of equality of (V_1) across data sets 383 could be rejected (comparison of bivariate models with homogeneous and heterogeneous V₁, P < 384 385 0.05, Fig. 3).

The among-individual between-trait correlations (r₁) reveal a broadly similar structure for the longand short-term studies (Table 3). Thus estimates for ST largely confirm our *a priori* expectation of positive correlation structure between the OFT traits and Emergence. One qualitative exception to the expected pattern is provided by Time in Middle. In LT this trait is positively correlated with all other traits as expected; however, in ST the sign of r₁ is negative (but not significant) between Time in Middle and Track Length and Emergence (Table 3).

Eigen decomposition confirms the view that qualitative differences between I_{LT} and I_{ST} are largely 392 393 related to Time in Middle. Thus, in both data sets the first eigenvector again dominates the variance in I (accounting for 66% and 73% in long- and short-term respectively), consistent with an important 394 395 latent character underlying behavioural variation (Fig. 2a). Time in Middle has a strong positive loading on EV1_{LT}, consistent with our *a priori* expectation that a bold fish would spend more time in 396 397 the middle of the open field arena, the corresponding loading coefficient is close to zero (in fact slightly negative) on EV1_{sT}. The angle (θ) between EV1_{LT} and EV1_{sT} is 34.63° (95% HPD interval, 398 5.03- 53.09°). While the point estimate of 34.63° indicates at least some divergence between the 399 leading eigenvectors on a scale from 0 (no difference) to 90 (axes are orthogonal), it is not 400 significantly greater than the angle expected by chance if the true matrices are identical (95% HPD of 401 the null distribution for θ generated by our parametric bootstrap is from 1.54 – 69.14°). While we 402 acknowledge that our null distribution indicates low statistical power to reject the null hypothesis 403 404 that θ = 0 (see Supplemental Fig. S3), our conclusion is however that EV1_{LT} and EV1_{ST} are broadly 405 similar, with qualitative differences largely attributable to the decreased loading of TIM on EV1_{ST}. This is further evidenced by a drop in θ from 34.63° to just 11.15° for the corresponding comparison 406 of I estimates excluding Time in Middle. There are also some qualitative inconsistencies evident 407 between EV2_{LT} and EV2_{sT} for the OFT traits, due to greater loadings on Track Length (changes sign), 408 Activity, Area Covered and Time in Middle, while the loading on Emergence is reduced (also changes 409 sign) (Fig. 2b). The angle (θ) between EV2LT and EV2ST = 48.32 ° (95% HPD interval 25.75-86.48 °), 410 411 which again is not significantly different from null expectations.

For those individuals tested in both long- and short-term studies, the among-individual correlations 412 between LT and ST data sets were positive (although not always significant based on likelihood ratio 413 tests) for OFT traits (Fig. 4), ranging from 0.219 (\pm 0.294) to 0.729 (\pm 0.314). Estimates were 414 significantly greater than zero for Area Covered and Time in Middle. However, we also found that 415 the correlation was significantly less than 1 for the traits Track Length and Activity. Thus, while 416 417 phenotypic performance of an individual in one data set may be predictive of its behaviour in the other, there is also evidence that the ranking of individuals, at least for Track Length and Activity, 418 419 significantly differs between long and short term studies. For Emergence the corresponding amongindividual correlation estimates between long- and short-term were actually negative, though not 420 421 significantly so. In fact the estimate was characterised by so much uncertainly that despite being 422 negative it was not possible to reject the null hypothesis of r = +1. We suggest this is a result of the 423 low repeatable variation of Emergence and thus little weight should be placed on this result.

424

425 Discussion

426 Data from our long-(LT) and short-term (ST) studies provide evidence of among-individual variance in behaviour, both when considered separately and in combination. Of the five traits assayed in the 427 428 two distinct types of behavioural trial - open field (OFT) and emergence and exploration (EET) -429 repeatabilities were statistically supported in all cases. In addition our analyses support the presence of a significant among-individual correlation structure for behavioural traits in I. 430 Correlation structure is found both within- and across-contexts (i.e. trial types), indicating 431 behavioural variation among fish that is consistent with accepted definitions of animal personality. 432 We found that repeatabilities of OFT traits were higher than the EET though not significantly so in all 433 cases. Our results therefore support the assertion of Burns (2008) that the OFT is a good and 434 435 reliable test of boldness and exploratory behaviour in small fish, although it is certainly possible that the EET could be better optimized to target the among-individual component. We discuss the 436

biological interpretation of (multivariate) variance within these two trial types further below.
However, here we note the pragmatic consideration that the binary distribution of Emergence data
obtained from the EET is more difficult to analyse and interpret while the censoring of Latency to
Emerge created a data distribution not readily modelled in any software. Although such problems
are likely surmountable by modification of the behavioural assay (e.g., using an extended
observation time to eliminate or at least reduce censoring), at least in this case it is not clear to us
that the EET provides additional biological insight.

444 Comparison of long- and short-term data sets suggested that the patterns of individual (co)variance between traits frequently used to define boldness are relatively stable. Nevertheless, as predicted a445 priori we found a tendency for the magnitude of R₁ to decrease with a higher interval between 446 observations, at least in OFT trials. For example, repeatabilities for OFT traits ranged from 0.188 to 447 448 0.458 in the short term data (with repeat observations at an average interval of four days) but 0.136 to 0.207 in the long term data (average interval of 56 days). In a meta-analysis of behavioural 449 450 repeatability studies that included either long- (i.e.> 1 year) or short-term (i.e. < 1 year) intervals between observations, the average (median) across all estimates was 0.37 (Bell et al. 2009). Here 451 our repeatability estimates pertain to correlated traits and are therefore not independent. 452 453 Nevertheless, apart from our short-term study estimates for Track Length and Activity, we note that our estimates for all other traits were lower than those of the meta-analysis average. Repeatability 454 455 estimates from short-term studies in the meta-analysis (Bell et al. 2009) outnumbered those from 456 long term studies by 11:1; however, our study considers observations collected within two distinctly separate periods across individual lifetimes. 457

Arguably the more important question to be asked of our long- and short-term data sets concerns
the stability of correlation structure within the multivariate I matrix and the interpretation of
boldness from its eigenvector decomposition. As seen with the single trait repeatabilities, the
structure of I_{LT} mirrored that of I estimated from all data combined. This is unsurprising given that

the long-term data comprised a much greater number of individuals and will thus dictate patterns in
the combined dataset. I_{LT} is dominated by a single vector that is broadly consistent with our
expectations of boldness. Significant within- and between- trial type correlations indicate that
individuals emerging from the EET refuge are more likely to have high scores for all OFT traits, thus
matching our expectation of bold behaviour.

Though not statistically significant, qualitative differences between I_{LT} and I_{ST} were apparent. These 467 differences were focussed around the sign and strength of correlations between Time in Middle and 468 469 traits from both trial types, indicating that both bold and shy individuals from the short-term study spent a similar amount of time in the middle, whereas in the long-term study, shy fish had behaved 470 in a more thigmotaxic manner. This pattern was reflected in comparisons of the major eigenvectors 471 of long- and short-term data, where a moderate, albeit not statistically significant, angle (θ) between 472 473 the first long- and short-term axes was estimated. Furthermore, if Time in Middle is dropped from the calculation, the estimated angle is reduced by more than half. Thus our interpretation is that 474 475 both data sets reveal a major vector of among-individual (co)variance in observed behavioural traits. This vector is similar in the two data sets and can be interpreted as a latent personality trait - namely 476 boldness. In both data sets bolder individuals tend to swim longer distances, be more active and 477 explore more area (in the OFT), and are more likely to emerge from a refuge (in the EET). However, 478 tendency to spend more time in the middle of the OFT arena appears not be a reliable indicator of 479 480 boldness as it was only associated with this vector in the LT study. Indeed this trait was the major 481 source of qualitative difference between the two matrices.

In the current study it is not possible to distinguish whether higher repeatabilities and the changing
structure of I with regard to Time in Middle are a consequence of the sampling period (long- vs.
short-term) or potentially reflect interesting, possibly even species-specific, biological changes that
happen with age and/or trial experience. Note, however, that our analyses control for any
habituation effects on mean behaviour, and that we found little statistical support for individual-by-

trial-number interactions (results not shown). More generally some authors have argued that 487 individual behaviour is likely to become more rigid and follow more set patterns over time (Roberts 488 and DelVecchio 2000). If so we would predict increasing repeatabilities with age (here confounded 489 490 with time scale of data collection). Conversely, others suggest that in the absence of any disturbance (e.g., in a constant laboratory environment), expectations of changes to individual 491 492 patterns of behaviour formed in early life are ill-founded (Stamps and Groothuis 2010). While no 493 overall differences were found between juvenile and adult behavioural repeatabilities in the Bell et 494 al. (2009) meta-analysis, a subset of data suggested juvenile behaviour to have higher repeatability than that of adults. However, the metanalysis contained only three studies that included 495 496 observations following individuals through from juvenile to adult status. Thus direct comparison of age classes is not straightforward. Clearly more empirical studies of how repeatability changes with 497 age would be valuable, as indeed would parallel studies exploring environmental dependence. Here 498 499 we assumed homogeneous variance structures across environments (density treatments, stacks) and other fixed effects (sexes, day order) for simplicity. These assumptions can be relaxed in the 500 statistical models to test for and quantify individual by environment (IxE) as changes in the among-501 individual variance (or structure of I in the multivariate case) (Dingemanse et al. 2010). Here post 502 503 hoc analyses of the LT data set provides some evidence of heterogeneous repeatabilities across 504 density treatment classes (see Supplemental Table S4). Though not expected to bias current 505 conclusions (parameter estimates presented are effectively averaged across treatments), if robust 506 this effect may certainly be biologically interesting.

The population level patterns of among-individual (co)variances between traits were broadly similar between LT₁ and ST₁, albeit with some differences as described above. However, by using the same individuals in both long- and short-term studies we were able to address the question of whether the relative ranking of individuals with respect to their behavioural tendencies was stable. The estimates of r₁ for each observed behavioural trait between the long- and short-term datasets provide a mixed answer to this question. Positive correlations for the OFT traits do show a degree of stability in (repeatable) behavioural tendencies across the data sets though statistical support was
mixed and it appears individuals were more likely to maintain a consistent ranking for some traits
(e.g. Area Covered) than others (e.g. Track Length).

We previously stated it is not our intention to be prescriptive about what boldness is or how it 516 517 should be assayed. Nevertheless, a priori, we anticipated that in the OFT, bold fish would travel long distances and be willing to visit a large area of the tank including the central zone, and that these 518 traits would correlate significantly with whether individuals emerged in the EET. However, this 519 520 depiction requires that the bold individual is also active and/or exploratory. Above we have noted that the major axis of variation in I is largely consistent with expectations of a bold-shy continuum as 521 522 the terminology is used in the literature; however, the strength of among-individual correlations suggests that it could equally be called exploration or general activity in a novel environment. 523 524 Nevertheless, as qualitatively almost all the variance loads onto this single axis of variance, we conclude that these continuums (personality axes) are, at least in our study species, either the same 525 526 entity or so tightly correlated that attempting to distinguish between them may have little practical value. Indeed, Burns (2008) concluded that emergence from a refuge was difficult to interpret 527 strictly as either boldness or exploration, even though it has been described as boldness only by 528 529 others, (e.g. Budaev 1997a; Brown et al. 2005). Exploring the functional significance of the consequences of this behavioural variance in wild populations is likely to yield more insight than 530 531 further debate with regard to terminology (e.g. Dingemanse et al. 2012a; Kurvers et al. 2012; 532 Carvalho et al. 2013). Nonetheless, we have sufficient statistical support in our results to conclude that both trial types revealed behaviours characteristic of boldness, evident from the strong among-533 individual correlations between all the observed traits. This again leads us in the direction of Burns' 534 (2008) view that in practice, the OFT offers the most useful test arena for this axis of personality. 535 Here we have obtained repeated measures of multiple behavioural traits during two test types and 536 537 across two distinct sampling periods (long- versus short-term), something that has seldom been 538 accomplished in the literature. In practical terms, we conclude that the open field trial is preferable

to the emergence and exploration trial as an experimental test for investigating boldness, and we 539 show how eigen decomposition of an I matrix can usefully identify latent personality traits. This 540 multivariate approach is broadly similar to that used in several other recent studies (Budaev 2010; 541 Carter et al. 2013; Araya-Ajoy and Dingemanse 2014). Our study also provides information about 542 the stability of personality, both in terms of population level patterns and individual differences. We 543 544 find that observed behavioural traits are repeatable over long time periods as well as when observations are made over only a few weeks, although there is a tendency for short term estimates 545 546 to be higher. Taking a multivariate approach we show that I is dominated by a single vector through phenotypic space that is similar across the two study periods and can be interpreted as boldness. 547 548 We note however, there are at least some qualitative differences in the relationships of observed behaviours to this vector. At the individual level we also find qualified support for the proposition 549 that short-term studies are informative for an individual's behavioural phenotype over longer (e.g. 550 lifetime) periods. 551

552

553 Ethical standards

554 Ethical review committees at the Universities of Edinburgh and Exeter approved all work in this 555 study, which was carried out under license granted by the Home Office (UK) under the Animals 556 (Scientific Procedures) Act 1986.

557

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a)

b

Zone 1 Zone 2 (middle)	Zone 2	Zone 1	 Refuge
	Zone Z		



b)







Tables

Table 1 Data set for long-term (LT) and short-term (ST) studies. Number (N) and sex of individuals involved: male (M), female (F), total (T). Periods of data collection and intervals between trial pairs. Number of trials conducted: OFT (Open Field Trial); EET (Emergence & Exploration Trial); $N_{LT} = 2448$, $N_{ST} = 320$. Mean age of fish in days at the start of each trial pair with standard error in parentheses; "-" indicates trial not performed

							Num	ber of					
Study		Ν		Data	collectio	n period	tria	als		Mea	an Fish Age	e (SE)	
						Days							
						between							
	М	F	т	Start	End	trials	OFT	EET	1	2	3	4	5
				May	May				203	259	372	427	
LT	223	150	373	2011	2012	56	1224	1224	(26.35)	(26.44)	(27.15)	(27.13)	-
				Feb	Feb				715	719	723	727	732
ST	16	16	32	2013	2013	4	160	160	(13.36)	(13.36)	(13.36)	(13.36)	(13.36)

Table 2 Behavioural traits recorded in OFT (Open Field Trials) and EET (Emergence and Exploration Trials)

Trial type	Measured trait		Definition
OFT	Track Length	(TL)	Distance swum (cm)
OFT	Activity	(Act)	Percentage of time moving at a minimum 1.5cm/sec (%)
OFT	Area Covered	(AC)	Area of tank floor covered (%)
OFT	Time in Middle	(TIM)	Time spent in Zone 2 (seconds, see Figure 1)
EET	Emergence	(Em)	Whether or not the fish emerged from the refuge (binary)

Table 3 Among-individual variance/covariance matrix (I) from the multivariate analysis of a) all data, b) long-term study and c) short-term study. Estimates of variance (V₁, diagonal) with among-individual between-trait covariances (COV₁) below the diagonal and among-individual between-trait correlations (r₁; above the diagonal). Standard errors are shown in parentheses for all parameter estimates. Traits: Track Length (TL), Activity (Act), Area Covered (AC), Time in Middle (TIM), Emergence (Em)

a) All Data	TL	Act	AC	TIM	Em
TL	0.130 (0.025)	0.865 (0.033)	0.750 (0.069)	0.162 (0.117)	0.304 (0.198)
Act	0.124 (0.024)	0.159 (0.026)	0.731 (0.065)	0.241 (0.106)	0.577 (0.182)
AC	0.097 (0.022)	0.104 (0.022)	0.128 (0.026)	0.653 (0.075)	0.414 (0.202)
ТІМ	0.026 (0.019)	0.042 (0.020)	0.102 (0.023)	0.192 (0.029)	0.540 (0.180)
Em	0.026 (0.017)	0.054 (0.018)	0.035 (0.018)	0.056 (0.019)	0.055 (0.024)
b) Long-term	TL	Act	Area	TIM	E
TL	0.143 (0.028)	0.892 (0.030)	0.777 (0.069)	0.238 (0.118)	0.272 (0.192)
Act	0.137 (0.026)	0.164 (0.028)	0.708 (0.072)	0.314 (0.106)	0.539 (0.180)
AC	0.108 (0.025)	0.106 (0.025)	0.136 (0.030)	0.704 (0.075)	0.458 (0.208)
ТІМ	0.041 (0.022)	0.058 (0.022)	0.118 (0.026)	0.207 (0.033)	0.607 (0.181)
Em	0.027 (0.020)	0.058 (0.020)	0.045 (0.021)	0.073 (0.022)	0.071 (0.028)
c) Short-term	TL	Act	Area	TIM	E
TL	0.458 (0.155)	0.926 (0.041)	0.640 (0.182)	-0.247 (0.256)	1.070 (0.513)
Act	0.381 (0.137)	0.369 (0.134)	0.812 (0.112)	0.017 (0.274)	1.001 (0.502)
AC	0.188 (0.095)	0.214 (0.097)	0.188 (0.089)	0.492 (0.222)	0.545 (0.524)
ТІМ	-0.083 (0.089)	0.005 (0.084)	0.106 (0.079)	0.248 (0.101)	-0.667 (0.557)
Em	0.165 (0.080)	0.139 (0.073)	0.054 (0.056)	-0.076 (0.059)	0.052 (0.066)

How stable are personalities? A multivariate view of behavioural variation over long and short timescales in the sheepshead swordtail, Xiphophorus birchmanni.

Behavioral Ecology and Sociobiology

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Supplemental Tables

Supplemental Table S1 MCMCglmm analyses of the binary Emergence trait. Table shows a) the intraclass correlation (IC - the binary equivalent of the repeatability (see methods)) from a univariate model, and b) among-individual correlation (r₁) estimates from bivariate models of Emergence and each open field trial trait

Model	Т	rait(s)	IC	rı	95% HPC) interval
					Lower	Upper
a)	Emergence	-	0.090		0.024	0.177
b)	Emergence	Track Length		0.641	0.303	0.999
	Emergence	Activity		0.736	0.488	0.977
	Emergence	Area Covered		0.560	0.308	0.920
	Emergence	Time in Middle		0.573	0.300	0.872

Supplemental Table S2 Univariate analyses of observed behavioural traits using the full (ALL), long- (LT) and short-term (ST) study data fitted using ASRemI. The among- (V₁) and within-individual (residual) variance (V_R) estimates are presented for each trait along with repeatability (R). χ^2_1 and P-values relate to likelihood ratio tests of the significance of V₁. Note that for univariate models only we assume the test statistic to be asymptotically distributed as a mix of 50:50 χ^2_0 and χ^2_1 (following Visscher 2006). Behavioural traits studied: Track- length (TL), Activity (Act), Area Covered (AC), Time in Middle of tank (TIM), Emergence (Em). Behavioural traits studied: Track Length (TL), Activity (Act), Area Covered (AC), Time in Middle of tank (TIM), Emergence (Em)

Data	Trait	Vi (SE)	V _R (SE)	R (SE)	χ ² ι	Р
	TL	0.132 (0.025)	0.658 (0.029)	0.167 (0.029)	56.10	<0.001
	Act	0.159 (0.027)	0.668 (0.029)	0.193 (0.029)	75.01	<0.001
ALL	AC	0.124 (0.026)	0.767 (0.033)	0.140 (0.027)	41.86	<0.001
	ТІМ	0.185 (0.029)	0.682 (0.030)	0.214 (0.029)	82.61	<0.001
	Em	0.058 (0.024)	0.889 (0.039)	0.061 (0.025)	6.88	0.005
	TL	0.143 (0.028)	0.689 (0.033)	0.172 (0.031)	41.51	<0.001
	Act	0.165 (0.028)	0.655 (0.031)	0.201 (0.031)	64.01	<0.001
LT	AC	0.141 (0.030)	0.768 (0.037)	0.155 (0.031)	31.76	<0.001
	ТІМ	0.206 (0.033)	0.693 (0.033)	0.229 (0.032)	69.85	<0.001
	Em	0.072 (0.028)	0.887 (0.043)	0.075 (0.029)	7.87	0.003
	TL	0.457 (0.154)	0.520 (0.067)	0.468 (0.093)	41.28	<0.001
	Act	0.369 (0.133)	0.571 (0.073)	0.393 (0.095)	29.23	<0.001
ST	AC	0.186 (0.089)	0.663 (0.085)	0.220 (0.089)	10.08	0.002
	ТІМ	0.248 (0.101)	0.594 (0.076)	0.295 (0.093)	17.20	<0.001
	Em	0.061 (0.069)	0.885 (0.113)	0.064 (0.071)	1.03	0.156

Supplemental Table S3 Estimates of fixed effects (with standard errors in parentheses) from univariate mixed models of each behavioural trait for the data combined and for the long- (LT) and short-term (ST) studies. Significance was assessed using conditional F statistics and all models contained a random effect of individual identity. Coefficients are not presented for Stack, Treatment and Trial*Treatment due to their being multilevel factors. All individuals from ST were housed in the same stack therefore this covariate was not included in the ST analyses. Traits: Track Length (TL), Activity (Act), Area Covered (AC), Time in Middle (TIM), Emergence (Em).

Dataset	Response	Fixed Effect	Coefficient (SE)	DF	F	Р
All	TL	Mean	1.898 (0.121)	1,332.7	1666.86	<0.001
		Sex	-0.046 (0.061)	1,348.8	0.56	0.454
		Day order	-0.006 (0.002)	1,1376.1	8.01	0.005
		Stack		6,545.5	53.11	<0.001
		Trial	0.230 (0.028)	1,1126.2	207.67	<0.001
		Treatment		3,339.3	1.56	0.201
		Trial*Treatment		3,1375.9	2.68	0.046
	Act	Mean	3.223 (0.125)	1,347.1	3860.25	<0.001
		Sex	-0.145 (0.064)	1,365.2	5.07	0.026
		Day order	-0.003 (0.002)	1,1373	1.96	0.164
		Stack		6,564.6	33.46	<0.001
		Trial	0.238 (0.028)	1,1129.5	225.4	<0.001
		Treatment		3,353.4	3.86	0.01
		Trial*Treatment		3,1374.5	4.75	0.003
	AC	Mean	2.795 (0.127)	1,339.8	2204.17	<.001
		Sex	0.252 (0.063)	1,354.5	15.76	<0.001
		Day order	-0.006 (0.002)	1,1376.9	7.57	0.006
		Stack		6,555.8	7.87	<0.001
		Trial	0.179 (0.030)	1,1141.5	112.28	<0.001
		Treatment		3,347.2	2.37	0.071
		Trial*Treatment		3,1363.8	2.24	0.083
	TIM	Mean	1.579 (0.128)	1,342.4	621.85	<0.001
		Sex	0.528 (0.067)	1,361.7	62.39	<0.001

		Day order	-0.009 (0.002)	1,1368.7	15.56	<0.001
		Stack		6,559.8	9.52	<0.001
		Trial	0.075 (0.029)	1,1119.7	10.36	0.001
		Treatment		3,348.2	0.85	0.47
		Trial*Treatment		3,1367.3	6.13	<0.001
	Em	Mean	0.665 (0.130)	1,297.7	141.35	<0.001
		Sex	0.222 (0.060)	1,301.2	13.54	<0.001
		Day order	0.007 (0.003)	1,1342	6.21	0.007
		Stack		6,525.1	9.53	<0.001
		Trial	-0.085 (0.032)	1,1138	8.31	0.004
		Treatment		3,306.7	2.63	0.051
		Trial*Treatment		3,1111.1	1.07	0.048
LT	TL	Mean	1.723 (0.144)	1,350.1	611.49	<0.001
		Sex	-0.043 (0.065)	1,348	0.44	0.505
		Day order	-0.007 (0.002)	1,1219.6	9.8	0.002
		Stack		5,353.6	7.81	<0.001
		Trial	0.310 (0.043)	1,976.1	226.07	<0.001
		Treatment		3,354.8	1.34	0.263
		Trial*Treatment		3,980.2	0.26	0.853
	Act	Mean	3.065 (0.143)	1,350.7	2106.73	<0.001
		Sex	-0.164 (0.066)	1,349.2	6.17	0.014
		Day order	-0.004 (0.002)	1,1219.5	3.16	0.078
		Stack		5,354.1	9.25	<0.001
		Trial	0.311 (0.042)	1,969.7	242.22	<0.001
		Treatment		3,355.2	2.89	0.036
		Trial*Treatment		3,973.6	0.67	0.571
	AC	Mean	2.707 (0.150)	1,341.7	1466.43	<0.001
		Sex	0.244 (0.067)	1,339.4	13.26	<0.001
		Day order	-0.007 (0.002)	1,1217.9	8.15	0.005
		Stack		5,345.4	5.21	<0.001
		Trial	0.282 (0.045)	1,973.9	99.55	<.001

		Treatment		3,346.5	1.32	0.27	
		Trial*Treatment		3,978.1	1.42	0.237	
	TIM	Mean	1.667 (0.150)	1,349.4	588.12	<0.001	
		Sex	0.540 (0.071)	1,348.4	58.39	<0.001	
		Day order	-0.010 (0.002)	1,1216.3	17.09	<0.001	
		Stack		5,352.6	8.49	<0.001	
		Trial	0.075 (0.043)	1,962.3	2.3	0.132	
		Treatment		3,353.7	1.27	0.285	
		Trial*Treatment		3,966.2	3.03	0.029	
	Em	Mean	0.654 (0.155)	1,336.3	143.76	<0.001	
		Sex	0.198 (0.064)	1,330.8	9.46	0.002	
		Day order	0.009 (0.003)	1,1179.3	7.51	0.007	
		Stack		5,342.4	4.67	<0.001	
		Trial	-0.085 (0.049)	1,983.7	7.9	0.005	
		Treatment		3,340.8	1.77	0.153	
		Trial*Treatment		3,995.8	1.39	0.244	
Т	TL	Mean	2.487 (0.508)	1,26.9	33.21	<0.001	
		Sex	0.064 (0.267)	1,38.2	0.06	0.81	
		Day order	0.013 (0.007)	1,122.2	4.11	0.046	
		Trial	-0.029 (0.065)	1,121	5.12	0.027	
		Treatment		3,27.2	0.72	0.547	
		Trial*Treatment		3,120.9	4.35	0.006	
	Act	Mean	3.144 (0.521)	1,26.9	69.66	<0.001	
		Sex	0.270 (0.253)	1,35.7	1.14	0.292	
		Day order	0.011 (0.007)	1,122.6	2.78	0.1	
		Trial	0.031 (0.068)	1,121.1	7.58	0.007	
		Treatment		3,27.1	0.99	0.411	
		Trial*Treatment		3,121	3.36	0.021	
	AC	Mean	1.263 (0.542)	1,27	17.17	<0.001	
		Sex	0.394 (0.212)	1,32.1	3.45	0.072	
		Day order	0.005 (0.007)	1,124.4	0.51	0.474	

		Trial	0.145 (0.073)	1,121.7	14.43	<0.001
		Treatment		3,27.1	2.82	0.058
		Trial*Treatment		3,121.5	0.65	0.588
	TIM	Mean	-0.027 (0.521)	1,27.1	3.99	0.056
		Sex	0.447 (0.224)	1,33.5	3.98	0.054
		Day order	0.001 (0.007)	1,123.6	0.01	0.904
		Trial	0.246 (0.069)	1,121.5	17.04	<0.001
		Treatment		3,27.2	2.87	0.056
		Trial*Treatment		3,121.4	1.46	0.231
	Em	Mean	1.534 (0.612)	1,26.9	19.29	<0.001
		Sex	0.438 (0.186)	1,28	5.53	0.026
		Day order	-0.005 (0.008)	1,135.4	0.43	0.512
		Trial	-0.062 (0.083)	1,125.7	0.73	0.395
		Treatment		3,27	1.69	0.193
		Trial*Treatment		3,123.1	0.79	0.502

Supplemental Table S4 Tests for heterogeneity of variance structures across density treatments for each behavioural trait in the LT data sets. Presented are χ^2 statistics with associated P-values for comparing models with homogeneous and heterogeneous (i.e. treatment specific) among-individual and residual variances. Significant heterogeneity of variance components across density treatments is indicated for Track Length (TL) and Time in Middle (TIM) only. Treatment specific variance components for TL estimated under the heterogeneous model (not shown) demonstrate lower repeatability (SE) in the High/High treatment R = 0.122 (0.075), relative to other treatment classes (Low/Low R = 0.311 (0.073), Low/High R = 0.209 (0.087), High/Low R = 0.273 (0.097)). For TIM, repeatability (SE) is reduced in the Low/High treatment R = 0.063 (0.037), relative to other treatment classes (Low/Low R = 0.212 (0.067), High/Low R = 0.206 (0.061))

Trait	χ ² 6	Р
Track Length	38.31	<0.001
Activity	7.18	0.30
Area Covered	1.99	0.92
Time in Middle	20.96	0.002
Emergence	2.07	0.91

How stable are personalities? A multivariate view of behavioural variation over long and short timescales in the sheepshead swordtail, Xiphophorus birchmanni.

Behavioral Ecology and Sociobiology

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Supplemental Figure Legends

Fig. S1 Summary of raw behavioural data showing observed mean (± standard error) by Trial in long- (dark grey) and short- (light grey) term studies for a) Track Length, b) Activity, c) Area covered, d) Time in middle, and e) Emergence, which is represented as a percentage and therefore does not have an associated error. The long-term study (LT) comprised four Trials, while there were five Trials in the short-term (ST) study

Fig. S2. Posterior distribution of the intra-class correlation for the binary trait of Emergence modelled in MCMCglmm. The posterior mode for the intraclass correlation, IC = 0.109, 95% HPD interval 0.041 – 0.194

Fig. S3 Parametric bootstrap distributions for θ , the estimated angle between EV1_{LT} and EV1_{ST} in the case that a) I_{LT} and I_{ST} are equal to their REML estimates, and b) I_{LT} and I_{ST} are equal such that the true angle between leading eigenvectors is zero. Distributions are based on 5000 pairs of simulated matrices (see main text for further details) a)



Track Length

b)



c)



d)

Time in Middle length of time (secs) LT ST Trial

e)



Fig. S2











