

1 **Sex differences in cognitive performance, style and domain relationships in**  
2 **mosquitofish (*Gambusia affinis*)**

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13

14 **Highlights**

15 - Males and females perform at similar levels in associative learning and cognitive  
16 flexibility assays, but females tend to outperform males on a spatio-temporal learning  
17 task.

18 - Female performance in associative learning trials (numerical discrimination task) can be  
19 predicted by cognitive style behaviors (exploration, reaction time, and activity); whereas  
20 male performance cannot.

21 - Males, but not females, show a predictive relationship between associative learning and  
22 cognitive flexibility performance.

23 - Our results demonstrate that sex differences in cognition extend beyond performance  
24 into cognitive style and domain relationships, suggesting that investigations into animal  
25 personality and cognition require more comprehensive characterization.

26

27 **Abstract**

28 Given that the sexes often differ in their ecological and sexual selection pressures, sex  
29 differences in cognitive properties are likely. While research on sexually dimorphic cognition  
30 often focuses on performance, it commonly overlooks how sexes diverge across multiple  
31 cognitive tasks (cognitive domains) and in behaviors associated with cognitive performance  
32 (cognitive style). We tested male and female western mosquitofish (*Gambusia affinis*) in three  
33 cognitive tasks: associative learning (numerical discrimination), cognitive flexibility (detour  
34 task), and spatio-temporal learning (shuttlebox). We characterized statistical relationships  
35 between cognitive performances and cognitive style during the associative learning task with  
36 measures of anxiety, boldness, exploration, reaction time, and activity. We found sex  
37 differences in performance, cognitive style, and the relationships between cognitive domains.  
38 Females outperformed males in spatio-temporal learning task, while the sexes performed  
39 equally in associate learning and cognitive flexibility assays. Females (but not males) exhibited  
40 a 'fast-exploratory' cognitive style during associative learning trials. Meanwhile, only males  
41 showed a significant positive relationship between domains (associative learning and cognitive  
42 flexibility). We propose that these sexually dimorphic cognitive traits result from strong sexual  
43 conflict in this taxon; and emphasize the need to explore suites of sex-specific cognitive traits  
44 and broader comparative work examining sexual selection and cognition.

45

46 **Keywords**

47 sex differences, cognitive style, numerical discrimination, cognitive flexibility, *poeciliidae*,  
48 *Gambusia affinis*

49

50 **1. Introduction**

51 Individuals vary in many cognitive attributes such as learning ability, style, and consistency in  
52 performance across varying tasks. These three attributes commonly referred to as cognitive  
53 performance, cognitive style, and cognitive domain respectively, are often examined  
54 individually, but recent efforts have begun to identify how they may be inter-related [1-3]. For  
55 instance, some of the early work in this arena has suggested that how quickly an animal  
56 makes a decision and how much it explores its environment (e.g. a fast-exploring cognitive  
57 style) should predict an individual's performance (accuracy) in a learning task [2]. Empirical  
58 studies have shown variable support for this hypothesis with some taxon revealing a positive  
59 relationship [4] between fast-exploratory styles and learning performance, others revealing a  
60 negative relationship [5-6], and yet other studies find that a fast-exploratory cognitive style is  
61 unrelated to performance [7]. Furthermore, relationships between cognitive style and  
62 performance are often domain-specific [8]. Understanding how and why these relationships  
63 vary across taxa and between domains is a current challenge in cognitive studies, however, a  
64 factor that is emerging as one of the predominant predictors of variation across cognition is sex  
65 [8].

66

67 A recent meta-analysis by Dougherty & Guillette (2018) identified sex as the single best  
68 categorical variable that explains the variation across animal studies on the associations  
69 between personality and cognition [8]. This is unsurprising given fitness benefits for particular  
70 cognitive traits are often sex-dependent [9-13], leading to differential selective pressures  
71 favoring certain cognitive traits in one sex and potentiating sexual dimorphism. While cognitive  
72 sex differences are accumulating across the literature, there exists a great deal of variation  
73 between and within taxa [14] and very little work examining how these sex differences bear out  
74 across cognitive domains. Sex differences in cognitive performance are frequently domain-  
75 specific [15-21], and even when learning performance is equivalent between the sexes the

76 behavioral predictors that underlie individual variation can differ [22-23]. The present study is  
77 designed to specifically examine sex differences in cognitive performance, cognitive style, and  
78 relationships between cognitive domains in a taxon with well-defined sexually dimorphic  
79 behaviors [24-27].

80

81 The *poeciliidae* family of livebearing freshwater fish— which includes the guppies, mollies,  
82 mosquitofish, and swordtails— exhibits a wealth of natural variation in sexual selection  
83 pressures [24] and cognition [25], making this family a uniquely suitable system to address sex  
84 differences in cognition [26]. Poeciliids are famous for sexually divergent reproductive roles as  
85 males provide only sperm via an intromittent organ (gonopodium) and females undergo a  
86 monthlong internal gestation period. Mosquitofish (*Gambusia* sp.) are characterized by a high  
87 degree of sexual conflict, with males exhibiting some of the highest rates of sexual harassment  
88 across the poeciliidae family and females adopting a strong avoidance responses including  
89 shoaling to reduce male harassment [27]. Artificial selection experiments on male gonopodial  
90 length in this genus results in larger female (but not male) brain size [28], suggesting that  
91 sexual conflict may act differentially on cognition. In our study we assessed the cognitive  
92 performance of the western mosquitofish (*Gambusia affinis*) in three cognitive domains:  
93 associative learning (numerical discrimination assay), cognitive flexibility (detour task), and  
94 spatio-temporal learning (shuttlebox assay). To evaluate cognitive style, we measured the  
95 relationship between associative learning performance and a variety of behaviors displayed  
96 during the numerosity discrimination test trials. We then examined whether performance in one  
97 cognitive domain predicted performance in another and whether these relationships varied by  
98 sex.

99

100 By examining sex differences in cognitive performance across associative learning, cognitive  
101 flexibility and spatio-temporal learning, we can begin to determine how divergent sexual  
102 selection pressures influence these cognitive domains. Based on previous studies with this  
103 species, we expect similar numerical discrimination performances between the sexes [22].  
104 Based on work with guppies, we predict that females will outperform males in a detour task  
105 [15,17]. Our spatio-temporal assay (shuttlebox) has only previously been performed on one  
106 other fish taxon (zebrafish [29]) with no sex-dependent effect. However, given the strong

107 selection pressures on females to find refuge from male harassment, we predicted that female  
108 *G. affinis* are likely to outperform male *G. affinis* in a spatio-temporal learning assay that varies  
109 time and place of a shoal group. Moreover, we predicted that the sexes would diverge in  
110 cognitive style where a ‘fast-exploratory’ learning type (e.g. faster decision making, shorter  
111 latencies to sample) would be more associated with males as has been found in other  
112 poeciliids [7]. Lastly, we predicted that the sexes would diverge in their relationship between  
113 cognitive domains. While the literature generally suggests a negative relationship between  
114 associative learning and cognitive flexibility [30], studies in poeciliids have thus far have not  
115 found a relationship [15,31]. Domain relationships between spatio-temporal learning and  
116 associative learning have not yet been explored in poeciliids, however, we hypothesized that  
117 *Gambusia affinis* females will exhibit a positive relationship between these two domains as  
118 both are critical to shoaling decisions.

119

120 **2. Materials and Methods**

121 **(a) Housing**

122 Wild-caught western mosquitofish *Gambusia affinis* (27 female, 27 male) from outdoor ponds  
123 at Brackenridge Field Laboratories in Austin, TX were group housed in 35 gallon aquaria at  
124 24.4-26.6°C on a 13-11 light cycle. Prior to testing, individuals were socially isolated for 2  
125 days in 2.5 gallon aquaria. Individuals participated in 13 days of cognitive assay testing with  
126 assay order balanced across individuals and sexes, and 24 hour interval between assays.

127

128 **(b) Numerical Discrimination Experimental Design**

129 To test associative learning individuals were placed in a modified 10-gallon automated  
130 numerical discrimination experimental tank for an 11 day assay including 2 days of habituation,  
131 6 days of training, and 3 days of testing (Supplementary Fig 1). Stimuli were geometric  
132 shapes (adapted from Etheredge et al 2018 and controlled for non-numerical cues) [22]  
133 presented on LCD screens attached to the end sides of each tank. Version control of the  
134 automation scripts are available at <https://github.com/jenkins-cummingslab/ethoStim>; and for a  
135 more detailed description of our automated numerical discrimination operation see  
136 *Supplementary Methods*.

137

138 Trials occurred five times daily during habituation and training. During habituation, a food  
139 reward was administered simultaneously at both blank screens for four minutes. During  
140 training individuals were presented two 1:2 ratios (5 versus 10 shapes or 6 versus 12 shapes)  
141 haphazardly alternating between left and right reward sides across training trials. Stimuli  
142 appeared on screens for an initial 10 seconds prior to a food reward descending into the tank  
143 for an additional 10 seconds. Half of the individuals received food reward on the side with the  
144 greater quantity on the screen, and half of the individuals received food reward on the lesser  
145 quantity side. Individuals were then tested (no reward) for three days three times daily on novel  
146 testing ratios of varying difficulties: 1:2 (7 vs 14 shapes), 2:3 (8 vs 12 shapes), and 3:4 (9 vs 12  
147 shapes), with reinforcement trials with original training ratios following each test trial to prevent  
148 extinction).

149

150 We assessed learning performance during the initial 20 seconds in which stimuli were  
151 presented on the screens which corresponded to the reward administration time period during  
152 training. Similar to other numerical discrimination studies [22,32], we employed a ‘learning  
153 criterion’ of individuals that had at least one testing ratio with a median performance value  
154 (proportion time spent in the closest quarter of the tank to the correct screen) above 60%.

155

### 156 **(c) Cognitive Flexibility (Detour maze) Experimental Design**

157 The experimental tank (filled to 13cm) was subdivided into a starting alley (26 x 14cm), center  
158 section (14 x 31cm), and reward section (14 x 31cm). Because female *Gambusia affinis* shoal  
159 with other females to reduce male harassment [27,33] we used a male social activator and a  
160 female social reward. The male was placed behind the starting alley in a visible container. The  
161 center section contained a 25cm wide glass barrier which prevented individuals travelling in a  
162 direct line from reaching the social reward. The reward section contained the female in a  
163 visible container. Individuals could solve the detour task by turning away from the direct line  
164 and travelling through the unobstructed zones (3cm) at each side of the glass barrier to reach  
165 the social reward (Supplementary Fig 2). The focal individual habituated for five minutes in an  
166 opaque tube then swam freely for 10 minutes. Motivation was recorded as the latency to  
167 reach the barrier, and solution speed was recorded as the time difference between arrival at  
168 the barrier and reaching the social reward.

169

170

171 **(d) Temporal learning (Shuttlebox) Experimental Design**

172 Our experimental tank was 52 x 26cm filled a depth of 10cm, with an Adafruit 7" LCD Display  
173 at either end. The focal individual swam freely during a 5-minute habituation in which both  
174 screens displayed a video of an empty tank. After habituation, one screen presented a 20  
175 second video of 5 conspecific females, followed by a 90 second inter-stimulus interval (ISI) of  
176 the empty tank on both screens, then by the conspecific video on the opposite side  
177 (Supplementary Table 1). This alternating stimuli-ISI pattern lasted one hour as in [29].

178

179 Learners were evaluated as those who spent >50% of interaction time (within 10cm of  
180 screens) in the correct region (screen with an imminent shoal group appearing) during the final  
181 30 seconds of the ISI for three consecutive trials following the 4<sup>th</sup> ISI. Individuals who became  
182 non-active before the fourth ISI were not included in the analysis.

183

184 **(e) Video Scoring**

185 Human observers scored time spent in regions of the numerical discrimination tank (using  
186 CowLog 3.0.2 and a python-generated grid overlay  
187 [https://github.com/kjw2539/make\\_a\\_grid\\_eagle.py](https://github.com/kjw2539/make_a_grid_eagle.py)). Additionally, we recorded latency to  
188 change regions following the image presentation (reaction time), number of unique zones  
189 visited (exploration), and total transits between zones (activity). One author (KW) and twelve  
190 undergraduate students independently scored 315 videos (Single Score Intraclass Correlation  
191 between eleven scorers  $p = 5.41 \times 10^{-90}$ ). Recordings for the Detour Maze and Shuttlebox  
192 assays were taken using Debut Video Capture Software and LifeCam cameras. Detour videos  
193 were independently scored by hand by two undergraduate student scorers and compared to  
194 the co-author (KW) scorer ( $p = 4.83 \times 10^{-45}$  and  $p = 2.03 \times 10^{-35}$ ). Shuttlebox videos were hand  
195 scored by three undergraduate student scorers using a python program developed by Luke  
196 Reding (see  
197 [https://github.com/lukereding/shuttlebox/blob/master/track/shuttlebox\\_hand\\_track.py](https://github.com/lukereding/shuttlebox/blob/master/track/shuttlebox_hand_track.py)) that  
198 evaluated the position of the fish at 5 second intervals.

199

200

201

202 **(f) Statistics**

203 We used multiple linear regressions to determine correlations between continuous variables.  
204 We conducted an unpaired t-test or unpaired Wilcoxon signed-rank test (determined by a  
205 Shapiro-Wilk normality test) for continuous data split into 2 categories, or a One-way ANOVA  
206 or Kruskal-Wallis test depending on normality for continuous data split into >2 categories. For  
207 categorical data we conducted Chi-squared tests on data with >5 observations per category  
208 and Fisher's Exact Test for data which had <5 observations in a category. Data analysis and  
209 visualization were conducted using RStudio (3.2.2). Data analysis coding scripts, original data,  
210 and protocols can be found at <https://github.com/kjw2539/Comparative-Cognition-R-scripts>).  
211

212

**3. Results**

213 54 individual *G. affinis* were run through a series of three cognitive assays ( numerosity  
214 discrimination, detour maze, and shuttlebox), however, due to technical errors not all  
215 individuals completed all assays. Completion tallies across assays varied with 36 individuals  
216 completing numerosity discrimination (18 males, 18 females); 52 individuals completing detour  
217 maze (25 males, 27 females), and 26 individuals completing shuttlebox (9 males, 17 females).  
218

219

**(a) Associative Learning Performance (Numerical Discrimination)**

220 In the numerical discrimination task, 23 individuals successfully met the learning criterion  
221 (12F/11M) and 13 individuals did not (6F/7M). Sex did not influence the distribution of learners  
222 to non-learners ( $\chi^2 = 0$ ;  $p = 1.00$ , Fig 1a). Our metric of performance correlated significantly  
223 with several others measured (Supplementary Fig 3; e.g. first side chosen ( $r^2 = 0.175$ ,  $p =$   
224 0.013), and latency to enter the correct side ( $r^2 = 0.349$ ,  $p = 0.0004$ )). There was no difference  
225 in performance between individuals trained to the higher quantity or lower quantity ( $p = 0.465$ )  
226 or across ratios ( $F = 0.302$ ,  $p = 0.7399$ ). The sexes did not differ significantly in their learning  
227 performance ( $t = 1.483$ ,  $p = 0.148$ , Supplementary Fig 5). Body size (standard length), a proxy  
228 for age in female *G. affinis* [34], did not predict learning performance (females:  $p = 0.719$ ,  
229 males:  $p = 0.770$ , combined:  $p = 0.589$ , see Table 1) and average standard length did not differ

230 between individuals classified as learners and non-learners (females:  $p = 0.494$ , males:  $p =$   
231 0.285, combined:  $p = 0.839$ , Table 1).

232 **(b) Behavioral Differences in Associative Learning (Numerical discrimination)**

233 In the numerical discrimination assay, males and females exhibited the same levels of  
234 exploration ( $W = 192$ ,  $p = 0.350$ ), reaction time ( $W = 121$ ,  $p = 0.203$ ), sociability ( $W = 202$ ,  $p =$   
235 0.214), boldness ( $t = 1.5621$ ,  $p = 0.128$ ) and activity ( $W = 217.5$ ,  $p = 0.082$ ). Males exhibited a  
236 higher proportion of time in regions associated with anxiety (thigmotaxis) than females ( $W =$   
237 97,  $p = 0.040$ , Supplementary Fig 4). Female standard length was significantly positively  
238 correlated with exploration ( $r = 0.539$ ,  $p = 0.021$ ) and higher activity ( $r = 0.464$ ,  $p = 0.052$ ), but  
239 negatively correlated with boldness ( $r = -0.471$ ,  $p = 0.048$ , see Table 1). Male standard length  
240 showed no significant relationship to any of these behaviors (Table 1).

241

242 **(c) Cognitive Style in Associative Learning (Numerical Discrimination)**

243 Males and females differed in their relationships between behavior and performance. Female  
244 learning performance was significantly correlated with reaction time ( $r = -0.541$ ,  $p = 0.025$ , Fig  
245 2a) and exploration ( $r = 0.510$ ,  $p = 0.036$ , Fig 2b), and marginally significant with activity  $r =$   
246 0.477, ( $p = 0.053$ , Fig 2c). Male behaviors did not relate to performance (reaction time:  $p =$   
247 0.689, exploration:  $p = 0.332$ , activity:  $p = 0.489$ , Fig 2d-f).

248

249 **(d) Cognitive Flexibility Performance (Detour Maze)**

250 In the detour maze, 27 individuals (14F/13M) successfully navigated around the transparent  
251 barrier, 25 did not (13F/12M) (Fig 1b). Of the 25 individuals who did not navigate around the  
252 barrier, 9 (4F/5M) did not participate (approach the barrier). Size did not differ between solvers  
253 and non-solvers (Table 1). On average it took solvers 80 seconds to solve the task upon  
254 reaching the barrier. Latency to solve the maze was not predicted by sex ( $W = 101$ ,  $p = 0.645$ ,  
255 Supplementary Fig 6), size (females:  $r = -0.358$ ,  $p = 0.344$ , males:  $r = -0.454$ ,  $p = 0.219$ ,  
256 combined:  $r = -0.156$ ,  $p = 0.537$ ), or motivation (latency to approach the barrier,  $r = -0.107$ ,  $p =$   
257 0.596, Supplementary Fig 6). Motivation did not differ by sex ( $W = 221.5$ ,  $p = 0.846$ ) or size  
258 (females:  $r = -0.217$ ,  $p = 0.438$ , males:  $r = 0.218$ ,  $p = 0.455$ , combined:  $r = -0.050$ ,  $p = 0.795$ )  
259 and did not influence whether an individual navigated around the barrier ( $W = 223.5$ ,  $p = 0.860$ ,  
260 Supplementary Fig 6).

261

262

263 **(e) Spatio-Temporal Learning Performance (Shuttlebox)**

264 Of the 54 individuals tested in the shuttlebox assay, three were removed due to technical  
265 errors, and fifteen were removed from analysis for nonparticipation. We identified 7 learners  
266 (all female) and 19 non-learners(9M/10F). A near-significant sex difference in performance  
267 was found, where females reached the learning criterion more often than males (Fisher's exact  
268 test,  $p = 0.058$ , Fig 1c).

269

270 **(f) Relationships in Performance across Domains**

271 Males who solved the detour maze exhibited significantly higher numerical discrimination  
272 performance than non-solvers ( $t = -2.361$ ,  $p = 0.035$ ; Fig 3a). This relationship was not found in  
273 females ( $t = -0.673$ ,  $p = 0.511$ ).

274

275 **4. Discussion**

276 We identified sex differences emerging across multiple attributes of cognition: in performance  
277 within a given task, in the cognitive styles that predict performance, and in the relationships in  
278 performance between cognitive tasks.

279

280 **(a) Sex Differences in Cognitive Performance**

281 Similar to findings in previous poeciliid studies utilizing numerical, color, and shape  
282 discrimination, we observed no sex differences in associative learning in our numerical  
283 discrimination assay [22,35]. We found similar ratios of learners to non-learners as other  
284 numerical discrimination experiments with *G. affinis* (two-thirds learners [22]). We found no sex  
285 difference in cognitive flexibility as measured via our detour task, which is contrary to findings  
286 in a related poeciliid (*Poecilia reticulata*) where females outperform males in this task [15].  
287 Confirming our prediction, we found a sex difference in performance in the spatio-temporal  
288 learning assay, with females being more likely to reach the learning criterion than males. The  
289 only previous examination of this cognitive task in teleosts showed no sex differences in  
290 zebrafish [29].

291 The greater spatio-temporal learning performance in female *G. affinis* may be driven by  
292 sexually dimorphic motivations to shoal. Females from poeciliid species with high levels of  
293 sexual coercion like *Gambusia* employ shoaling as a strategy to reduce male harassment[27],  
294 and *Gambusia* females shoal more than males [22,36]. Female *G. affinis* have been shown to  
295 choose shoal group size in a rational manner [37]; and the benefit for selecting larger shoals  
296 can lead to both reduced male harassment and increased foraging efficiency in a closely  
297 related poeciliid (*Gambusia holbrookii*) [38]. Increased shoaling tendencies in females may  
298 drive the shoaling-related cognitive advantage observed in *G. affinis* females as seen in our  
299 spatio-temporal learning task in which females were more successful than males in predicting  
300 the time and place of the shoal. This sexual dimorphism in shoaling— and the sexual conflict  
301 that drives it— may manifest in sex differences in spatio-temporal cognitive tasks that emulate  
302 shoaling decisions.

303

304 **(b) Sex Differences in Cognitive Style**

305 Performance in numerosity discrimination tasks was predicted by a suite of related behaviors  
306 exhibited during the test trials by females but not males. Female mosquitofish that exhibited a  
307 ‘fast-exploratory’ cognitive style in the numerical discrimination assay exhibited higher  
308 associative learning performance. Specifically, females that explored a greater area of the  
309 tank, reacted faster, and moved more demonstrated higher numerical discrimination  
310 performances. This ‘fast-exploratory’ type observed in these mosquitofish females appears to  
311 share attributes of previously described ‘fast behavioral’ type individuals (more exploratory,  
312 active, bold, and aggressive [2]) in other taxa. A positive association between ‘fast behavioral’  
313 type and associative learning is documented across multiple taxa particularly in response to  
314 predation [8], including black-capped chickadees [439], sticklebacks [40], and Panamanian  
315 bishop fish (*Brachyraphis episcopi*) [41]. The ecological pressures that might lead to  
316 individuals adopting fast behavioral types may stem from being exposed to threatening  
317 environments (high predation), which places a selective pressure on the speed at which they  
318 sample their environment.

319

320 In some populations of guppies, intense male harassment has driven females into different  
321 habitats with greater predation levels than males [42]. Have the intense social pressures found

322 in *Gambusia* driven sex-specific habitat differentiation and thus shaped female fast-exploratory  
323 cognitive styles? In previous fish studies examining sex differences in cognitive styles during  
324 associative learning assays with food rewards (T maze [43], visual discrimination of shapes  
325 and colors [44]) researchers have found a fast-exploratory cognitive style associated with  
326 males, not females. Only during shoal discrimination tasks in poeciliids (guppies, *Gambusia*)  
327 have females been shown to make faster decisions than males [7]. If strong sexual selection  
328 pressures have shaped female *Gambusia* shoaling decision-making processes, perhaps this  
329 has influenced the cognitive style in general numerical discrimination tasks.

330

331 The cognitive style that is identified in females during the numerosity discrimination task in this  
332 experiment is comprised of behaviors that tend to covary by female size (exploration, boldness  
333 and activity, Table 1). Poeciliid females experience indeterminant growth, and therefore  
334 standard length is often a proxy for female age. While larger/older females were no better at  
335 numerosity discrimination than smaller females, they did trend towards being more exploratory,  
336 reacting faster and moving more in the test trials (Table 1). This reflects findings showing that  
337 larger females tend to disperse farther [45]. Are female *G. affinis* developing a cognitive style  
338 as they age? Age-dependent decision-making processes have been documented among  
339 female poeciliids. For instance, relative to smaller females, larger and presumably older female  
340 El Abra swordtails show stronger preference for courting phenotypes over coercive  
341 phenotypes [46] and exhibit less transitivity in mate choice decisions involving male size [47].  
342 Whether the increased exposure to male harassment over a female's lifetime contributes to the  
343 development of fast-explorative female cognitive style can only be determined with  
344 manipulative social experiments.

345

346 It is imperative to caution that the causal relationship between cognitive style and cognitive  
347 performance is unknown [2,8]. In addition to implying that our 'high learning individuals' could  
348 be the result of individual differences in activity and exploratory tendencies, we must  
349 acknowledge that the behavior observed in the numerical discrimination testing trials could be  
350 a result of cognitive performance—i.e. those individuals who quickly learn the task may more  
351 quickly habituate and thus become more active and exploratory. Further studies assessing the

352 developmental sources of variation in cognitive style and cognitive performance would help us  
353 understand the causal relationship between behavior and cognitive ability.

354

355

356 **(c) Sex Differences Across Cognitive Domains**

357 In addition to performance and cognitive style, the sexes differed in their cross-domain  
358 relationships of performance: males (but not females) who successfully solved the cognitive  
359 flexibility task (detour maze) showed significantly higher associative learning performance.  
360 This relationship has not previously been described in poeciliid fish. Given that acquisition and  
361 reversal learning utilize different neural mechanisms [48-50], our finding of a positive  
362 relationship in male performance across these domains suggests non-modularity in male  
363 cognition, or a potential “g-factor” of general intelligence [51]. A general intelligence factor  
364 typically describes roughly 40% of individual variation in human tests, and g -factors are often  
365 found in studies across multiple animal taxa including birds, mice, chimpanzees, and dogs  
366 [14]. But what might explain our sex differences in the cross-domain relationships? *Gambusia*  
367 are extremely invasive (found in over 40 countries [52]) and thus frequently experience highly  
368 variable environments. They likely must utilize both associative learning *and* cognitive flexibility  
369 as a strategy to succeed in these dynamic environments. This concept, known as the “adaptive  
370 flexibility hypothesis,” emphasizes that cognitive flexibility is an adaptive response to a  
371 changing physical [53] or social [54] environment. An investment in a positive domain  
372 relationship between associative learning would predominantly benefit *Gambusia* males given  
373 that males are more likely to disperse than females [45] and disperse farther [55]. However,  
374 whether this sex-specific domain relationship is driven solely by potential ecological differences  
375 between the sexes or some contribution of different sexual selection pressures is yet to be  
376 determined.

377

378 **4. Conclusion**

379 Our study identified new sex differences in spatio-temporal learning, sex-specific cognitive  
380 styles in associative learning, and a sex-specific positive relationship between performance  
381 across cognitive domains. A wealth of literature has identified sex differences in cognitive  
382 performance in mammals [56], birds [18], reptiles [19], and fish [25-26]. But here we find that

383 sex differences extend beyond performance into other cognitive attributes such as cognitive  
384 style and cross-domain relationship suggesting that more comprehensive characterization of  
385 cognition is important. Our experimental design in which the same individuals were assessed  
386 for cognitive performance and style across domains allowed us to find previously undescribed  
387 sex differences in cognition in *Gambusia affinis*. Fish exhibit a wealth of sex-specific ecological  
388 [20,57] and sexual selection pressures [24,26-27], therefore we can expect fish to continue to  
389 be an insightful taxonomic group in uncovering predictive patterns of sex differences in  
390 cognition [26]. Further studies, particularly those utilizing more extensive suites of cognitive  
391 testing, investigating neural mechanisms, and identifying developmental basis of these  
392 relationships will be critical to elucidate mechanisms governing the patterns observed in this  
393 study. In addition, comparisons of related species that differ in degree of sexual conflict and  
394 ecological pressures will be an important next step in distinguishing the factors that drive  
395 individual variation in cognitive performance.

396

### 397 **Data Accessibility**

398 Raw data (Microsoft Excel) as well as analysis documentation are available at the following  
399 archived Github repository: <https://github.com/kjw2539/Comparative-Cognition-R-scripts>

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### 411 **Compliance with ethical standards**

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414 The authors have no competing interests.

415 **Ethical approval**

416 The authors certify that this work followed ethical treatment of animals outlined in their IACUC  
417 protocol (AUP-2016- 00246).

418 **Authors' contributions**

419 Molly Cummings and Kelly Wallace conceived of the study. Kelly Wallace, Mary Ramsey, and  
420 Richie Rausch designed and constructed the experimental setup and data collection  
421 procedures. Kelly Wallace collected the data and performed statistical analyses. Kelly Wallace  
422 and Molly Cummings interpreted the results and wrote the manuscript. All authors gave final  
423 approval or publication.

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577 **Fig 1. Learning performance differs between the sexes in the shuttlebox assay, but not**  
578 **the numerical discrimination or detour tasks.** In the numerical discrimination assay, the  
579 sexes show equal distributions of learners and non-learners (learners reaching a minimum  
580 median performance of 60% for any of the 3 test ratios (7:14, 8 vs 12, 9 vs 12)) **(a)**. Detour  
581 maze solvers versus non-solvers also show equal distributions across the sexes **(b)**. Females  
582 reach the learning criterion in the shuttlebox assay (learners exhibiting three consecutive trials  
583 in which a majority of interaction time during the last 30 seconds of the ISI was spent within  
584 10cm of the correct screen) more than males (Fisher's Exact Test,  $p = 0.058$ ) **(c)**.

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586

587 **Fig 2. The sexes differ in cognitive style in the numerical discrimination task.** Female  
588 learning performance (proportion time spent in contingency zone during initial 20 sec of test  
589 trial) is significantly predicted by female reaction time ( $p = 0.025$ )**(a)** , exploration ( $p = 0.036$ )  
590 **(b)**, and activity (marginal significance,  $p = 0.053$ ) **(c)** displayed during test trials; whereas no  
591 significant correlations were found between these measures in males **(d-f)**.

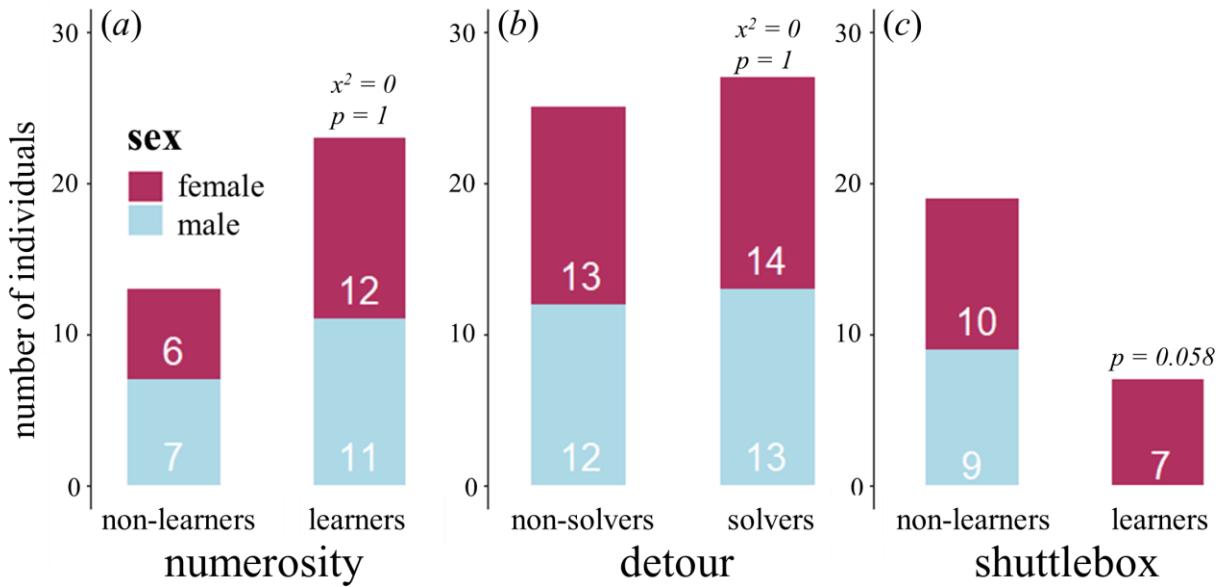
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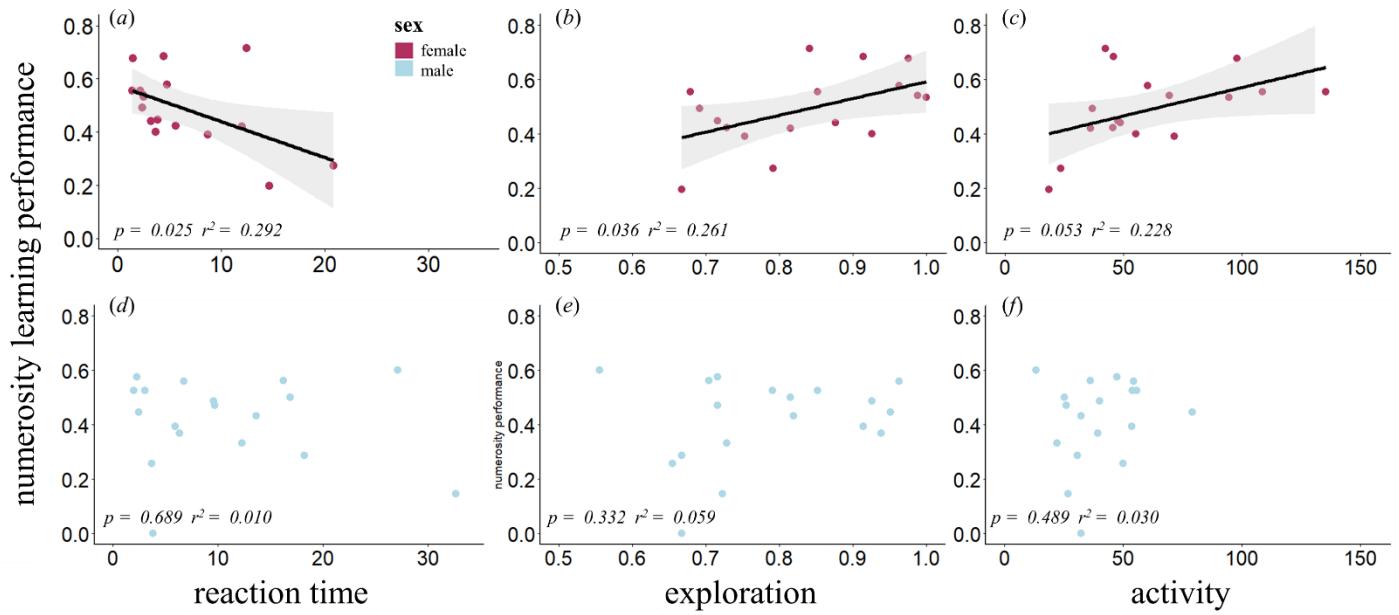
593 **Fig 3. Male performance is predicted across cognitive domains.** Males who solved the  
594 cognitive flexibility task (detour maze) exhibited significantly higher associative learning  
595 (numerical discrimination) performance than males who did not solve the cognitive flexibility  
596 task ( $p = 0.035$ ) **(a)**. Latency to detour around the barrier once it was approached did not differ  
597 by sex and did not predict associative learning (numerical discrimination) performance **(b)**.

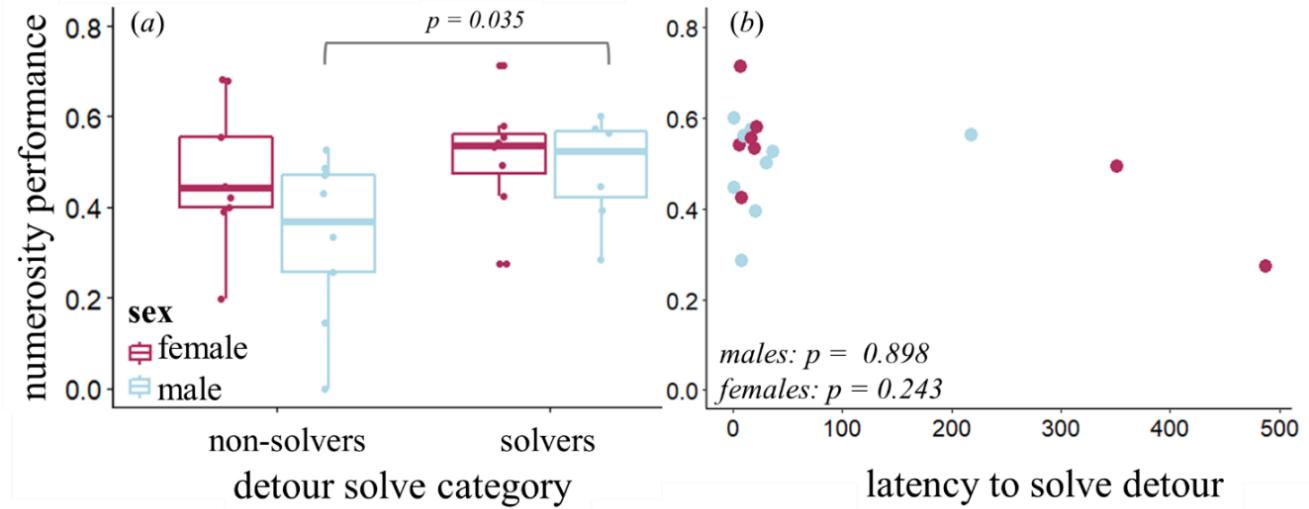
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600 **Table 1. Female size predicts cognitive style behaviors in the numerical discrimination**  
601 **task.** The relationship between standard length (a proxy for age) and the behavior listed is  
602 shown for three datasets: only females, only males, and all individuals combined. Continuous  
603 behavioral data results were determined using a multiple linear regression, and the reported  
604 effect size is Pearson's correlation coefficient. Categorical behavioral data results (denoted  
605 with \*) were determined using a t-test, and the reported effect size is Cohen's D. Significant p-  
606 values are highlighted in red.







Behavior	females p	females effect size	males p	males effect size	both sexes p	both sexes effect size
Numerical Discrimination Performance	0.719	-0.094	0.770	-0.074	0.589	-0.095
Numerical Discrimination Learning Category*	0.687	-0.204	0.595	0.267	0.839	-0.068
Detour Solve Category*	0.641	0.225	0.139	0.770	0.284	0.364
Detour Latency to Solve	0.344	-0.358	0.219	-0.454	0.537	-0.156
Detour Motivation	0.438	-0.217	0.455	0.218	0.795	-0.050
Shuttlebox Learning Category*	0.434	6.676	na	na	<b>0.034</b>	5.504
Numerical Discrimination Exploration	<b>0.021</b>	0.539	0.182	0.329	<b>0.009</b>	0.430
Numerical Discrimination Boldness	<b>0.048</b>	-0.471	0.706	-0.096	0.423	-0.138
Numerical Discrimination Anxiety	0.234	0.295	0.733	-0.086	0.434	-0.135
Numerical Discrimination Activity	0.052	0.464	0.122	0.378	<b>0.0008</b>	0.535
Numerical Discrimination Reaction Time	0.144	-0.358	0.148	-0.355	<b>0.024</b>	-0.376

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