1 Behavioural and life-history responses of mosquitofish to

2 biologically-inspired and interactive robotic predators

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Giovanni Polverino^{1,2}, Mert Karakaya³, Chiara Spinello³, Vrishin R. Soman³, Maurizio
Porfiri^{3,4,*}

- 6 7
- 8 ¹Centre for Evolutionary Biology, University of Western Australia, Perth, Australia
- ⁹ ²Department of Biology and Ecology of Fishes, Leibniz-Institute of Freshwater Ecology and
- 10 Inland Fisheries, Berlin, Germany
- ³Department of Mechanical and Aerospace Engineering, Tandon School of Engineering, New
- 12 York University, Brooklyn, United States
- ⁴Department of Biomedical Engineering, Tandon School of Engineering, New York
 University, Brooklyn, United States
- 15
- 16
- 17 *Author for correspondence:
- 18 Maurizio Porfiri
- 19 mporfiri@nyu.edu

20 ABSTRACT

Invasive alien species threaten biodiversity worldwide and contribute to biotic 21 22 homogenization, especially in freshwaters where the ability of native animals to disperse is 23 limited. Robotics may offer a promising tool to address this compelling problem, but whether and how invasive species can be negatively affected by robotic stimuli is an open 24 25 question. Here, we explore the possibility of modulating behavioural and life-history 26 responses of mosquitofish by varying the degree of biomimicry of a robotic predator, whose appearance and locomotion are inspired by natural mosquitofish predators. Our results 27 28 support the prediction that real-time interactions at varying swimming speeds evoke a more 29 robust antipredator response in mosquitofish than simpler movement patterns by the 30 robot, especially in individuals with better body conditions that are less prone to take risks. Through an information-theoretic analysis of animal-robot interactions, we offer evidence in 31 32 favour of a causal link between the motion of the robotic predator and a fish antipredator 33 response. Remarkably, we observe that even a brief exposure to the robotic predator of fifteen minutes per week is sufficient to erode energy reserves and compromise the body 34 condition of mosquitofish, opening the door for future endeavours to control mosquitofish 35 in the wild. 36

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38 Keywords

Animal personality; bioengineering; biomimetics; body condition; invasive species;predation risk

41 **1. INTRODUCTION**

The presence of animal species in areas where they are not native is common across the globe, with tremendous costs for both human activities and the ecological integrity of those areas [1, 2]. Despite efforts from both governmental and academic institutions, existing methods for eradicating invasive alien species (IAS) or mitigating their negative effects remain labour-intensive, economically unviable, and, often, ineffective [3].

Freshwater animals are particularly vulnerable to IAS, whereby native species are confined to smaller water bodies and their ability to disperse is limited compared to other ecosystems [4]. Mosquitofish (*Gambusia affinis*, Baird and Girard, and *Gambusia holbrooki*, Girard) are among the most widely diffused freshwater IAS in the globe, and their negative impact on indigenous animal communities (via aggressive behaviours and/or predation [5-8]) has been recognized by the International Union for Conservation of Nature that listed mosquitofish among the world's hundred worst IAS [9].

54 Technical efforts to eradicate mosquitofish from water bodies and mitigate their negative impact on the native fauna are, however, limited. For example, increasing the structural 55 complexity of the environment through artificial refugia was successful in reducing mortality 56 in barrens topminnow (Fundulus julisia, Williams and Etnier) exposed to mosquitofish under 57 laboratory settings, but beneficial effects from artificial refugia disappeared in the wild [10]. 58 Similarly, the use of fish toxicants to combat the spread of invasive mosquitofish resulted in 59 detrimental consequences for native fish [11]. The utilization of floating traps to target 60 61 mosquitofish near the water surface has been shown to be a successful technique, but it is a labour intensive process that can be pursued only in small sites and for short periods of time 62 [12]. 63

Robotics constitutes a promising tool for addressing some of these challenges, by 64 65 offering a versatile, customizable, and consistent approach to modulate the behavioural response of live animals [13-15]. Particularly relevant are experiments that have shown the 66 67 possibility of eliciting behavioural responses in freshwater fish through biologically-inspired robots triggering a cost-benefit decision process [16-21]. The use of robotics in the study of 68 predator-prey interactions might afford the design of new hypotheses-driven studies that 69 could unfold the basis of fear and anxiety in prey fish [22-25] and illuminate the 70 71 evolutionary consequences of nonlethal exposure to predators [26, 27]. Just as robotics might bring new scientific insight into predator-prey interactions, it also contributes to
 ethics in animal experimentation by minimizing potential harm to live animals.

74 In particular, previous research efforts from our group indicate that a robotic fish can be 75 designed to repel mosquitofish [28] and simultaneously attract non-invasive fish under 76 laboratory settings [19]. The possibility to isolate fish from one species to another allows 77 safeguarding non-invasive species from the aggressive attitudes of mosquitofish, thereby 78 providing compelling evidence for the use of biologically-inspired robots as a possible method for the selective control of mosquitofish in the wild. However, the technology to 79 80 deploy autonomous robotic fish in a complex ecological environment to control the 81 behaviour of mosquitofish is still in its infancy, calling for a scientifically-principled 82 understanding of how mosquitofish interact with biologically-inspired robotic stimuli.

Mosquitofish can adjust their behavioural and life-history strategies in response to 83 varying environmental conditions, especially in the attempt to minimize risks of predation 84 85 [29]. Mosquitofish are typically less prone to take risks [30] and invest less in reproduction [31] and energy reserves [32] under predation risk than in more beneficial conditions, with 86 plastic adjustments associated with predation risks that can eventually result in the whole 87 body morphology of mosquitofish to be reshaped [33]. Visual cues represent the 88 89 predominant factor for predator recognition in most freshwater fish [34], especially 90 mosquitofish [28, 35], and a growing literature has provided convincing evidence that visual 91 cues from animated images [36-38] and biologically-inspired robots [19, 28] can be used to 92 influence mosquitofish behaviour.

93 While experiments comparing mosquitofish behavioural response to computer-animated and robotic stimuli are presently lacking, evidence from other freshwater fish suggest that 94 visual stimuli associated with a biologically-inspired robotic predator might elicit a stronger 95 96 response than computer-animated images [23]. Experiments in [23] have compared the fear response of zebrafish (Danio rerio, Hamilton) evoked by live predator fish, a robotic replica 97 98 of the predator fish, and computer-animated images of the predator fish, determining that: (a) the robot caused a robust avoidance response in zebrafish that was comparable to that 99 observed for live predators, while computer-animated images did not, and (b) individual 100 101 responses were more consistent over time when zebrafish were exposed to the robot than 102 to live predators and computer-animated images. In addition to these methodological 103 observations, practical considerations toward future deployment in the wild favour the use of robots over computer-animated images. In fact, practicality challenges the use of computer-animated images in the wild, where it may be unfeasible to employ computer screens or projectors. Based on these methodological and practical aspects, we favour the use of robotic stimuli in place of computer-animated images.

108 In this study, we sought to test whether behavioural and life-history responses of 109 mosquitofish could be modulated through a robotic predator whose visual appearance and locomotion were inspired by mosquitofish predators, the largemouth bass (Micropterus 110 111 salmoides, Lacépède; Figure 1). Largemouth bass coexist with mosquitofish in the wild and 112 constitute their most common predators [39, 40], with mosquitofish representing over 80% 113 of the fish consumed by juvenile largemouth bass in their native environments [41]. Our 114 biologically-inspired robotic predator was designed to take advantage of the innate antipredator behaviour that largemouth bass induce on mosquitofish under laboratory 115 settings [31, 42]. We repeatedly exposed mosquitofish to robotic predators varying in their 116 degree of biomimicry to disentangle the relative contribution of the robot swimming and its 117 interactivity on both behavioural and life-history adjustments associated with antipredator 118 responses in mosquitofish. We hypothesized that: (a) visual stimuli from the robotic 119 predator would repel mosquitofish, as suggested in [28, 36], (b) increasing the degree of 120 121 biomimicry in the motion of the robot would increase antipredator behaviours and impact life-history strategies (that is, energy reserves) in mosquitofish, and (c) individuals would 122 differ from each other in the extent of their antipredator responses [43], with individuals 123 124 with high future expectations (that is, individuals with high energy reserves) being more risk-averse than others [44]. 125

From a methodological point of view, our study contributes to the state-of-the-art in 126 animal-robot interactions [13, 14, 45] along several research directions. First, we established 127 128 a robotic platform that allows for tailoring the degree of complexity of the interaction through a closed-loop control system, integrating real-time tracking and high-precision 129 130 robotics. Through this platform, we successfully varied the degree of biomimicry of the interactive robotic predator, by simulating random attacks toward the fish at either 131 constant or increasing speed. This experimental manipulation effectively allowed for the 132 quantification of the relative contributions of typical locomotory patterns of predators in 133 triggering antipredator responses in mosquitofish. Second, we shied away from a rigid 134 135 prototype, in favour of a soft robotic replica that incorporates a spine-like structure to

promote natural oscillation that are reminiscent of body undulations, which are known to 136 be critical for fish-robot interactions in the water [18, 21]. Third, we integrated traditional 137 means of behavioural analysis with modern elements of dynamical systems theory, through 138 139 the information-theoretic framework of transfer entropy [46]. Through the lens of transfer 140 entropy, we demonstrated an improved comprehension of the antipredator response of 141 mosquitofish, by testing for potential cause-and-effect relationships between the motion of the robotic predator and mosquitofish antipredator response. Finally, although few recent 142 143 studies have considered behavioural response of animals repeatedly confronted with robots 144 [47, 48], a detailed study of individual variation in mosquitofish behaviour was lacking, 145 especially in the context of life-history consequences of the exposure to robotic stimuli.

146 Although focused on mosquitofish, the theoretical and methodological underpinnings of this work could inform research on other IAS, whose presence in the environment is also a 147 threat to biodiversity and economy. For example, recent studies have demonstrated the 148 possibility of utilizing robots inspired by live predators to influence the behaviour of locusts 149 (Locusta migratoria, Linnaeus), a major pest for human agricultural economies and 150 151 ecosystems stability [49, 50]. Similarly, the peregrine falcon-like robot Robird has been recently presented for deployment in the aviation industry to deter birds from flying in the 152 153 vicinity of aircrafts [51].

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155 2. MATERIALS AND METHODS

156 **2.1 Study organism and maintenance**

A total of 150 wild-caught Western mosquitofish (*Gambusia affinis*, Baird and Girard) were purchased from a commercial supplier (Carolina Biological Supply Co., Burlington, NC, USA) and were acclimatized for one day in stock tanks. Then, 75 focal individuals (average body length of 2.9 \pm 0.3 cm) were randomly selected from stock tanks, with sick individuals and fish showing physical and/or behavioural anomalies excluded *a priori*.

Focal fish were housed individually in transparent Plexiglas cylinders (10 cm diameter), placed within a large housing tank (185 x 47 x 60 cm, length, width, and height) and submerged in water per 10 cm, as in [29, 52]. The lateral surface of the transparent cylinders was perforated to promote water circulation across separate cylinders, affording visual and chemical interaction among individuals despite physical isolation. This housing scheme prevented aggression, competition for resources, and sexual harassment among 168 mosquitofish, with each cylinder marked with a unique identification code to facilitate the 169 identification of individuals over time. The position of the cylinders was periodically 170 randomized to allow visual and chemical interactions among all fish. Fish were acclimatized 171 in the cylinders for one month before experiments, and they were housed in these cylinders 172 for the whole duration of the study (approximately three months).

Fish were kept under a 12h light/12h dark photoperiod and fed with commercial flake food (Nutrafin max; Hagen Corp., Mansfield, MA, USA) once a day. Water parameters were checked daily, with temperature and pH maintained at 26°C and 7.2 pH, respectively, throughout the study.

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178 **2.2 Experimental setup**

179 2.2.1 Experimental arena for behavioural tests

180 Behavioural trials were performed in an experimental arena ($44 \times 30 \times 30$ cm, length, width, and height), filled with 10 cm of conditioned water (Figure 1A). The walls and the bottom 181 surface of the arena were covered with white opaque contact paper to control for external 182 183 disturbance and optimize automated computer tracking of fish motion during trials. Two 38 184 W fluorescent tubes (All-Glass Aquarium, UK) were mounted 130 cm above ground and 185 were used to provide homogeneous illumination to the apparatus. A high-resolution 186 webcam (Logitech C920 webcam, Lausanne, Switzerland) was mounted 140 cm above the 187 floor for a complete overview of the experimental arena.

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189 2.2.2 Robotic platform and predator replica

190 The experimental arena was supported by aluminium T-slotted bars 29 cm above the ground to allow the placement of the robotic platform underneath (Figure 1A). The platform 191 192 allowed for manoeuvring the robotic replica along the three degrees of freedom: two degree of freedom were controlled for in-plane translational motion of the replica and one 193 degree of freedom served to adjust the predator body rotation. The replica was 194 magnetically connected to the platform through a 3D-printed base made of Polylactic Acid 195 196 filaments (3.2 cm x 1.0 cm x 0.6 cm length, width, and height) containing two circular 197 neodymium magnets (0.63 cm thick and 0.3 cm diameter) and an acrylic rod (4 cm length and 0.62 cm diameter; Figures 1A, 1B, and 1C). The in-plane translational motion was based 198 on a Cartesian plotter (XY Plotter Robot Kit, Makeblock Co., Ltd, Shenzhen, China) and the 199

body rotation was controlled via a stepper motor (NEMA 14, Pololu Corp., Las Vegas, NV,
USA). Further details on the robotic platform are in the Supplementary Material. The
platform was originally designed in [53] to study zebrafish social behaviour and utilized in
[54] to examine zebrafish learning.

204 Locomotory patterns of the predator replica were inspired by pilot tests performed on 205 three juvenile largemouth bass (7.0 ± 0.5 cm), purchased from Teichwirtschaften Armin Kittner in Quitzdorf am See, Germany (<u>https://www.teichwirtschaft-kittner.de/</u>), before the 206 beginning of the experiment (Figure 1D). Live bass were placed individually in the 207 208 experimental arena and their behaviour was recorded over 30 minutes. Swimming 209 trajectories and swimming speeds were then obtained through an offline tracking software 210 developed by our group [55]. Mean and maximum swimming speed measured in the pilot tests and a swimming trajectory representative of the bass behaviour in the experimental 211 arena were used for the motion of the predator replica. 212

213 The morphology and coloration of the replica were also chosen to capture salient features of juvenile largemouth bass (Figures 1B, 1C, and 1D). Toward this aim, we took 214 photos of the live bass from different angles and estimated their body dimensions using a 215 dedicated software (ImageJ, National Institute of Health, Bethesda, Maryland, USA). The 216 217 body morphology of the replica was accordingly modelled in Solidworks (Dassault Systèmes SolidWorks Corp., Waltham, Massachusetts, USA) to create a 3D design and, then, a solid 218 mould. A spine-like structure in Polylactic Acid filament material was 3D-printed and 219 220 integrated within the 3D printed mould of the predator replica together with two glass eyes, relatively smaller than in live bass (Figure 1B). Then, the mould was filled with non-toxic and 221 aquarium safe silicone (Dragon Skin 10 Medium, Smooth-On, Macungie, PA, USA) and let 222 dry. The spine-like structure provided support to the weight of the silicone body of the 223 224 replica and facilitated body oscillations during swimming. Lastly, the silicone body of the replica was hand-painted using non-toxic, aquarium safe, and silicone based light grey and 225 silver paints (Smooth-On, Inc., Macungie, PA, USA) to mimic the characteristic coloration 226 pattern of largemouth bass (Figure 1C). Colour reflectance comparison between live bass 227 and its robotic replica were not performed. However, non-toxic pigments utilised to paint 228 the body of the robotic replica have been shown to be effectively perceived as natural 229 pigments in bluefin killifish (Lucania goodie, Jordan; [20]), a freshwater fish with well-230 developed vision like mosquitofish. 231

The moulded silicone body with glass eyes and spine-like structure was attached to a clear acrylic rod, connected to 3D-printed base with magnets. The clear acrylic rod allowed for setting the swimming depth of the biologically-inspired predator replica in the middle of the water column, that is, where the antipredator response of mosquitofish is known to be the strongest [28].

237

238 2.2.3 Experimental conditions and live tracking

We designed a series of experimental conditions with robotic replicas varying their motion 239 240 to proxy different degrees of biomimicry of live predators. In one control condition, the 241 experimental fish were tested in the absence of the replica (no predator: NP). In a second 242 control condition, the replica was motionless and positioned randomly within the arena before each trial started (predator motionless: PM). In the four experimental conditions 243 244 where a swimming replica was employed, the replica swam on either the predetermined 245 trajectory inspired by live bass (open-loop: OL) or it alternated between the predetermined trajectory and targeted real-time interactions (closed-loop: CL) with the focal fish. In two OL 246 conditions, the biologically-inspired predator replica followed the predetermined swimming 247 trajectory, either at a varying speed based on the motion of the live predator (OL1) or at a 248 249 constant speed (OL2). In condition OL2, the trajectory from the live bass was processed to manoeuvre the replica at a constant speed. Specifically, we locally fitted the trajectory using 250 cubic splines (interparc, Copyright (c) 2012 John D'Errico) and placed equally-spaced 251 252 waypoints on the splines such that the replica would move at a constant speed. The constant speed was chosen to be 6 cm/s to match the mean speed observed in juvenile 253 largemouth bass in our pilot tests and provide a dynamically rich visual stimulation for 254 mosquitofish. The same speed was used as the mean value of the speed profile in condition 255 256 OL1, consistently scaling experimental observations.

In the CL conditions, the replica, besides following swimming trajectories at a varying speed, was programmed to interact in real-time with the focal fish and to perform simulated attacks at random. However, the replica always performed an attack every minute of the trial for a total of 15 attacks. During an attack, the replica either accelerated to attain a large speed (20 cm/s; CL1) comparable to the maximum speed of live bass attacking a prey [56], or swam at a constant speed toward the fish (6 cm/s; CL2). When the replica was commanded to attack the focal fish, its motion was a function of the distance from the fish.

For CL1 condition, if the distance between the fish and replica was less than 1 cm, the 264 replica would only change its heading towards the direction of the focal fish and return to 265 the original heading; for distances between 1 and 10 cm (inspection zone in [57]), the 266 replica would change its heading, accelerate towards the fish at 20 cm/s², and stop at 267 approximately 1 cm from it; and for distances larger than 10 cm, the replica would change 268 the heading, accelerate at 20 cm/s² until reaching a speed of 20 cm/s, and maintain this 269 270 speed until stopping at 1 cm from the fish. For CL2 condition, if the distance between the 271 fish and replica was less than 1 cm, the replica would only change its heading towards the 272 direction of the focal fish and return to the original heading. For any distance greater than 1 273 cm, the replica would change its heading, and attack the fish with a constant speed of 6 274 cm/s and stop at 1 cm from the fish.

After an attack was completed, the replica returned to its original position prior to the 275 276 attack and restarted swimming along the predetermined trajectory until the next attack. 277 Notably, the region in which the robotic replica swam was smaller than the actual size of the experimental arena to allow at least 1 cm from the extremities of the replica's body (that is, 278 head and caudal fin) and the edges of the arena. This tolerance permitted smooth operation 279 of the robotic platform and avoided collision with the walls of the arena. Further details on 280 281 the real-time tracking system implemented for CL conditions are in the Supplementary 282 Material.

The custom-made software was calibrated on the exact size of an individual fish at each 283 trial (week) separately and utilized to calculate the following quantities: distance moved 284 (cm), time spent freezing (s), speed variance during swimming (cm²/s²), mean distance from 285 the predator replica (cm), predator inspection (counts), and time spent within one-body 286 length from the wall (s) – that is, thigmotaxis [58]. In particular, if a fish moved at a speed 287 288 less than half of its body length per second for two consecutive seconds, it was considered as freezing [59]. Predator inspection was estimated according to standard protocols 289 290 developed for guppies (*Poecilia reticulata*, Peters; [57]), a poeciliid species closely related to mosquitofish. In particular, we counted the number of events that a fish voluntarily 291 approached the predator replica by entering the 10 cm region around the replica while 292 actively swimming in its direction, that is, at an angle lower than ± 90 degrees from the 293 replica's head [57]. The distance from the wall used to estimate thigmotaxis was selected 294 295 based on pilot tests in which mosquitofish were exposed to the same robotic predator replica utilized in this study. Details on data extraction and tracking system are in theSupplementary Material.

Notably, reduced activity (in the form of short travelled distances and prolonged freezing) and large number of predator inspections, hesitancy in exploring open spaces that are unfamiliar and potentially dangerous (that is, high thigmotaxis), and erratic swimming patterns dominated by high speed variance are typically associated to risk aversion and fearful states in animals [43], including mosquitofish [28, 29, 36, 52, 59].

303

304 **2.3 Experimental procedure**

Once a week over seven consecutive weeks, fish were anesthetized in a solution of tricaine methanesulfonate (MS-222; 168 mg per 1L H₂O), sexed, and their body length (to the nearest 0.5 mm) and body weight (to the nearest 0.01 g) were measured. These measurements were conducted before the experiment started (baseline body measurements) and after the conclusion of each behavioural trial (week 1 to week 6). The Fulton's condition factor *K* (weight length⁻³ 10⁴, g mm⁻³ 10⁴; [60]) was then calculated as an index for the nutritional state (that is, body condition) of each fish at each week.

In each trial, a mosquitofish was gently hand-netted and placed into an opaque cylinder 312 313 in the experimental arena for 5 minutes to allow acclimatization to the setup. During acclimatization, the motors of the robotic platform were turned off and fish had no visual 314 315 contact with the apparatus outside the opaque cylinder. Then, the opaque cylinder was gently removed and the platform turned on, allowing the fish to explore the arena in either 316 absence (NP) or presence of the biologically-inspired predator replica (PM, OL1, OL2, CL1, 317 and CL2 conditions) for 15 minutes. After the trial was completed, the fish was transferred 318 back into its individual housing cylinder and the next trial was initiated. 319

320 The behaviour of each individual (n=75) was tested once a week over six consecutive 321 weeks, with individuals tested once per condition. No mortality was reported such that an equal number of replicates were conducted for each condition. One week interval between 322 two consecutive behavioural measurements is commonly adopted when testing individual 323 variation in mosquitofish behaviour to minimize memory effects [29, 52]. Experimental 324 conditions were performed in a randomized order, but the NP condition was always 325 performed last to mitigate bias on fish baseline behaviour caused by individuals being 326 327 exposed to diverse degrees of predator threat, as observed in [42] for risk avoidance in mosquitofish. Fish were tested in a randomized order to exclude consistent differences intheir behavioural outcome caused by hunger [61].

330

331 **2.4 Statistical analysis**

We initially tested whether body length, body mass, and Fulton's K were correlated by 332 333 estimating phenotypic correlations (that is, the overall correlation attributable to betweenand within-individual correlations) with bivariate linear mixed-effects models (LMMs), as 334 suggested by [62]. In these models, we specified the individual as the random effect (that is, 335 336 random intercepts) to account for repeated measures of the same individual across weeks. Body size was correlated with both mass and K, while mass and K were not correlated with 337 338 each other (Table S1). Therefore, we included both body mass and Fulton's K as fixed effects in the LMMs below, while body size was excluded from the models. 339

Since we were interested in testing whether mosquitofish antipredator response 340 341 increased with an increased degree of biomimicry of the replica, we measured individual behaviour repeatedly across experimental conditions. We ran separate LMMs in which 342 distance moved, freezing, speed variance, mean distance from the replica, predator 343 inspection, and thigmotaxis were included one-by-one as the dependent variables. In each 344 345 model, individual identities were included as the random effect, while body mass, Fulton's K, sex, week, and condition (that is, the degree of biomimicry of the robotic predator) were 346 entered as fixed effects. A significant effect of condition in a given model (or any other fixed 347 effect included in that model) would indicate that condition explained a significant portion 348 of the behavioural variance observed after accounting for the variation explained by the 349 other fixed effects. The significance of individual differences was tested using both 350 likelihood ratio tests (LRTs) and Akaike information criteria (AICs), where a full model 351 352 including individual as a random effect was compared with a reduced model in which the 353 random effect was excluded. Random intercepts represented the proportion of the total 354 phenotypic variance not attributable to fixed effects that was explained by among-individual 355 variance, that is, differences in personality traits among individuals.

Building upon our previous work [15], we implemented the information-theoretic notion of transfer entropy to quantify the influence of the biologically-inspired predator replica on the behaviour of the live fish and *vice versa*. Given two stochastic processes, transfer entropy quantifies the reduction in the uncertainty in the prediction of the future of one of

the processes from its present due to additional knowledge about the other stochastic 360 processes [63]. In this vein, a nonzero value of transfer entropy indicates a potential 361 influence between the two processes [63]. Here, transfer entropy was computed on the 362 363 time-series of the speed of the replica and the mosquitofish, which were first down sampled 364 to 1 Hz to ensure that one time-step (1 s) would suffice to encode the response time of the 365 fish to the replica and vice versa. Therefore, a total of 904 points (904 s) were used per each trial. Then, we converted the time series into symbols depending on whether the speed 366 increased or decreased between two consecutive time-steps [64]. In agreement with [15], 367 368 we computed the transfer entropy from the replica (R) to the fish (F) as follows:

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$$TE_{Robot \to Fish} = \sum_{F_{t+1}, F_t, R_t} \Pr(F_{t+1}, F_t, R_t) \log_2 \frac{\Pr(F_{t+1}/F_t, R_t)}{\Pr(F_{t+1}/F_t)}$$

where *F* and *R* are the down-sampled time-series of the speeds, and Pr represents the probability mass function computed via plug-in estimation. By flipping *F* with *R*, we computed transfer entropy from the fish to the replica $TE_{Fish \rightarrow Robot}$. Across the five experimental conditions in which the predator replica was employed, transfer entropy could only be used in OL1, CL1, and CL2, since the speed of the replica was constant in PM and OL2 and, therefore, encoded no meaningful information.

We expected information flow in OL1 to be one-directional, since the replica swam 376 irrespective of the fish, which should be influenced by the swimming pattern of the replica. 377 On the other hand, the information flow in CL1 and CL2 was expected to be two-directional, 378 379 with the fish responding to the replica and the replica adjusting its attacks as a function of 380 the behaviour of the fish. For each of the three conditions (OL1, CL1, and CL2), we obtained 381 surrogate data from all the possible shuffling (74×74) of the identities of the fish and the replica within each condition. For each of these shuffling, we randomly selected 74 values 382 without repetitions to obtain a mean transfer entropy value; this process was repeated 383 20000 times to obtain a surrogate distribution. To ascertain an influence through transfer 384 entropy, we tested whether the corresponding experimental value was in the right tail of 385 386 the distribution. This process was conducted six times, twice for each of the three conditions to examine information flow in either direction (fish to robot or robot to fish, 387 388 similar to [65]).

389 We then tested whether transfer entropy differed across conditions and between 390 directions ($TE_{Fish \rightarrow Robot}$ and $TE_{Robot \rightarrow Fish}$). Therefore, we built an LMM with transfer entropy as the dependent variable, the direction of information flow, condition, and their interaction as fixed effects, and both individual identities (fish identity) and pair identities (fish and replica identities) included as random intercepts. As for the LMMs on behavioural traits described above, the significance of random intercepts (both individual and pair identities) was tested using LRTs and AICs.

396 Lastly, we were interested in testing whether fish energy reserves (Fulton's K) varied in response to the exposure to robotic predator replicas. Toward this aim, we built an LMM 397 with Fulton's K as the dependent variable, including individuals' identity as the random 398 399 effect (that is, random intercepts) and sex, week, and condition (that is, the degree of 400 biomimicry of the robotic predator) as fixed effects. We then tested whether the 401 behavioural variation observed across conditions reflected variation in Fulton's K. Based on our initial hypothesis and findings from behavioural analyses, experimental conditions were 402 403 consolidated in three categories: controls (K measured before the experiment started, after 404 tests performed in absence of the predator replica, and after the exposure to the predator motionless; baseline, NP, and PM, respectively), low degree of biomimicry (OL1, OL2, and 405 406 CL2), and high degree of biomimicry (CL1). Variation in K was then tested with an LMM, in which Fulton's K was the dependent variable, individuals' identities the random effect (that 407 408 is, random intercepts), and sex, weeks, and condition category the fixed effects.

409 Data analyses were performed in R version 3.5.1 [66] using the "Ime4", "nIme", "ImerTest", and "MCMCgImm" packages [67-70], estimated marginal means (EMMs) based 410 on univariate models and post hoc comparisons were performed with "emmeans" adjusted 411 412 for simultaneous inference with the *mvt* method [71], while permutation tests for transfer entropy analysis were conducted in MATLAB (R2018a; MathWorks, Natick, MA, USA [72]). 413 Prior to all analyses, speed variance was log-transformed to normalize error distribution in 414 415 the model's residuals. Except for the permutation test that is independent by error distributions, we assumed Gaussian error distributions that were confirmed for all response 416 417 variables after visual inspection of model residuals. The significance level was set at α < 0.05. 418

419

420 **3. RESULTS**

421 Behaviour was strongly dependent on the experimental condition in which mosquitofish 422 were tested after controlling for variation explained by week (see results from the LMMs in Table 1). The distance between the fish and the replica decreased when the replica was allowed to swim in the arena with respect to the condition PM where it was held in place (P<0.001 in pairwise comparisons between PM and any other experimental condition; Figure S1). This was especially evident when attacks were performed in real-time by accelerating toward the fish and the interactive nature of the replica buffered fish' attempts to be away from it (P<0.001 in pairwise comparisons between CL1 and OL1, OL2, and CL2; Figure S1).

On the contrary, fish tendency to inspect the predator replica did not vary across swimming replicas, that is, the number of inspections in CL1 was undistinguishable from OL1, OL2, and CL2 (Figure 2A). Accordingly, fish swam on average longer distances, varied their swimming speed more, and froze less when exposed to a swimming replica than in control conditions (*P*<0.001 in pairwise comparisons between NP and PM confronted with any other experimental condition, except for speed variance and freezing between NP and CL1 and between PM and CL1, respectively; Figure S1).

Thigmotaxis increased with increasing biomimicry in the replica's motion, whereby the 437 time interval spent in the proximity of the walls was longer when fish were exposed to a 438 replica varying its attacking speed in real-time (CL1) than other replicas (P<0.001 in pairwise 439 440 comparisons between CL1 and any other experimental condition in which a robotic replica was employed), with the shortest time observed in the presence of the motionless replica 441 442 (PM; Figure 2B). On the other hand, behavioural responses of fish exposed to an attacking 443 replica that swam at a constant speed (CL2) were comparable with those observed in openloop conditions (OL1 and OL2), consistently across all measured traits (Figures 2 and S1). 444

The variation in body condition (Fulton's K) among individuals was a significant predictor 445 for the variation in their behavioural response across conditions (see results from the LMMs 446 447 in Table 1). In particular, individuals with more energy reserves varied their swimming speed more (that is, exhibited higher speed variance) in response to the replica and an analogous 448 449 role of K was also noted, albeit not significant, with respect to distance moved, distance from the replica, predator inspection, and thigmotaxis (Table 1). Accordingly, individuals 450 with higher K tended to swim longer distances, maintained larger distances from the 451 replicas, inspected the replicas less, and spent more time in the proximity of the walls. 452 Nevertheless, we registered consistent among-individual variance in all traits after that 453 454 behavioural variation explained by the model predictors was accounted for, that is, fish differed in personality traits (see results from the LMMs in Table 2), except for the meandistance from the replica and the individual intercepts for the transfer entropy.

We failed to identify an information transfer flow in the open-loop condition OL1 in both 457 458 directions (that is, from the robot to the fish and vice versa; Figures 3A and 3B). On the 459 contrary, a significant information transfer was observed in both directions in CL1 (Figures 460 3C and 3D) and CL2 (Figures 3E and 3F). When comparing information transfers within conditions, we observed that transfer entropy from the robot to the fish in the open-loop 461 condition OL1 was higher than from the fish to the robot (P=0.003; Figure 3G), in agreement 462 463 with our expectations on the one-directional nature of the interaction in OL1. Transfer 464 entropy in the closed-loop condition CL1 was also larger from the robot to the fish than in 465 the opposite direction (P<0.001), while transfer entropy in CL2 was comparable between directions (Figure 3G). Importantly, the effect of the replica on fish behaviour was stronger 466 in CL1 than in CL2 (P=0.042; Figure 3G), while other pairwise comparisons were not 467 468 significant. In other words, the biologically-inspired robotic predator interacting with mosquitofish in real-time and accelerating toward the fish (CL1) was more effective in 469 eliciting antipredator responses in mosquitofish than when it attacked at a constant speed 470 (CL2). 471

We also found that body condition (Fulton's K) varied across experimental conditions 472 (see results from the LMM in Table S2), with K significantly lower after fish faced the 473 predator replicas than after fish were tested in absence of the replica (P<0.001 in pairwise 474 475 comparisons between NP and any other experimental condition; Figure S2). The decrease in K after exposure to the replica (P<0.001 in pairwise comparisons between controls and 476 replicas with either low and high biomimicry; Figure 4) appeared, however, to be 477 independent of the degree of biomimicry of the replica (non-significant pairwise comparison 478 479 between low and high biomimicry; Figure 4).

480

481 **4. DISCUSSION**

Here, we have disentangled the relative contributions of swimming pattern and closed-loop control of an interactive robotic predator on the antipredator behavioural response and lifehistory strategies in mosquitofish. Fish thigmotaxis increased with the degree of biomimicry in the motion of the replica, suggesting that integrating real-time feedback from mosquitofish position in the control of a replica interacting at increasing speed plays a key

role in eliciting antipredator response in mosquitofish. The quantification of the information 487 flow between the replica and fish supported the existence of a causal relationship between 488 fish antipredator response and the motion of the biologically-inspired replica. We also 489 490 observed that individual behaviour was relatively predictable, with variations in energy reserves explaining a large portion of the behavioural variance observed among 491 492 mosquitofish. Notably, energy reserves decreased after fish were exposed to the 493 biologically-inspired robot only 15 minutes per week, but variation in energy reserves did 494 not depend on the degree of biomimicry in the motion of the replica.

495 After the initial detection of a potential predator, a fish typically identifies and assesses 496 the threat based on cues from its natural predators [73]. The extent of an antipredator 497 response is determined from the trade-off between minimizing risk of predation and energy consumption toward survival and reproduction [74], such that greater threats produce 498 499 stronger avoidance [75]. Here, we provide experimental evidence that swimming patterns 500 represent a salient source of information for predator recognition in mosquitofish that regulate the extent of their antipredator response. This evidence is based on highly-501 controllable experiments that employ a state-of-the-art robotic predator replica, whose 502 visual appearance and swimming pattern were inspired by measurements on juvenile 503 504 largemouth bass, the main predator of mosquitofish in the wild [39-41]. Not only did the robotic replica allowed for controlling the swimming speed and acceleration of the predator 505 stimulus, but also it afforded the implementation of controlled attacks toward mosquitofish 506 507 to study their antipredator response in real time. By opting for a robotics-based platform in lieu of a live predator, we were able to exclude potential correlations between antipredator 508 response of mosquitofish and inherent biological variations in the predator behaviour (that 509 is, idiosyncrasies with focal individuals, fatigue, and hunger) that could confound 510 511 hypothesis-testing.

The more robust antipredator behavioural response was registered when mosquitofish were exposed to a replica swimming at a varying speed and performing targeted, fast attacks. Reducing the degree of biomimicry toward a replica that performed attacks in realtime at a constant speed resulted into a weaker antipredator behavioural response, similar to that registered with non-interactive replicas that followed predetermined swimming trajectories. This evidence aligns with prediction from the literature positing that speed and acceleration should play a key role on prey-predator interactions in fish [76]. Our

information-theoretic analysis of the interaction between the robotic replica and the fish suggests the presence of a cause-effect relationship underlying the antipredator behavioural response of mosquitofish, which confirms the expected link between a predator's attacking speed and fish behavioural response [77]. More specifically, we determined that the uncertainty in the prediction of the future speed of mosquitofish from its present speed was reduced due to additional knowledge about the speed of the replica, such that the motion of the replica encoded valuable information about the behaviour of mosquitofish.

Beyond the analysis of the mean behavioural response at the population level, we 526 527 discovered that a relatively short exposure to the biologically-inspired robotic predator (only 528 15 minutes per week) resulted in a substantial reduction in the whole body condition of 529 mosquitofish (index of fat reserves for a given body size; condition factor K) that did not depend on the swimming pattern of the robot. Recent evidence from multiple populations 530 of mosquitofish in the wild has shown that the condition factor K in mosquitofish decreased 531 532 on average of 5.8% over a five months period in response to severe environmental challenges associated with water pollution [78]. Here, we observed that the body condition 533 declined of 3.1% over a week, after mosquitofish were exposed once to a predator replica, 534 thereby suggesting a hidden effect of the robot on mosquitofish life-history adjustments. 535 536 This finding aligns with evidence of nonlethal effects of predator-prey interactions [26], whereby costs of antipredator responses extend to ecologically-relevant traits beyond 537 behaviour, such as physiology and body condition [79]. 538

539 In fact, theory predicts that stress responses affect the way animals allocate resources to fuel emergency functions [79], with animals investing relatively more energy in survival (that 540 is, escaping from the predator) and relatively less in future reproduction (that is, energy 541 reserves) with increasing predation risk [80]. With respect to mosquitofish, nonlethal effects 542 543 of predator exposure have been found to lower their body condition, ultimately leading to lower fertility and fecundity rates [31]. Under this perspective, evidence from this study 544 545 indicates that a relatively brief exposure to a biologically-inspired robotic predator compromised the body condition of mosquitofish. Notably, the body condition increased 546 again when mosquitofish were tested in the arena in the absence of the replica, indicating 547 that variation in body condition resulted from the exposure to the robotic predator than 548 other factors (for example, time, exposure to the arena, and handling of the fish). 549

550 At the individual level, we found that fish differed consistently from each other in the extent of their antipredator response across six repeated exposures to robotic predators 551 varying in their degree of biomimicry (that is, fish differed in personality traits; [43]). While 552 553 the presence of personality variation among mosquitofish is well documented in the 554 literature (see for example [29, 52, 59] and references therein), this study offers evidence 555 that meaningful variation in antipredator response among mosquitofish can be successfully captured using robotic stimuli. Interestingly, a large portion of the variance in the 556 antipredator response observed among mosquitofish was explained by variation in their 557 558 body condition. In particular, individuals in better body conditions varied their swimming 559 speed more in response to the robotic predator, tended to swim longer distances, 560 maintained larger distances and inspected less the replicas, and spent more time in the proximity of the wall than mosquitofish in poorer body conditions. Individuals can trade-off 561 survival at the cost of future reproduction, but the antipredator behavioural response of an 562 563 individual should also depend on its body condition [81] as the reproductive value is condition-dependent. In this vein, our results are in agreement with predictions from the 564 life-history theory that individuals with high future expectations (that is, individuals with 565 high energy reserves) should systematically be more risk-averse than others [81]. Therefore, 566 567 our findings suggest that antipredator behavioural response toward robotic predator fish differs at the individual level in a relatively predictable manner. 568

This study contributes to the state of the art on the modulation of the behaviour of 569 570 invasive and pest species through the use of predator-mimicking robotic fish [19, 28], supporting the technological evolution of pest control agents, along similar line of 571 development as insects [50] and birds [51]. Specifically, we aimed at the precise 572 quantification of granular features of predator locomotion on antipredator responses of 573 574 invasive mosquitofish through the development of a state-of-the-art robotic predator whose swimming characteristics can be controlled across a continuum range of biomimicry. 575 576 Our findings build on previous research efforts on the modulation of mosquitofish behaviour through biologically-inspired robots, shedding light on the role of the robot 577 morphology on mosquitofish behaviour [28] and addressing the differential response of 578 mosquitofish and zebrafish to robots [19]. In particular, we demonstrated that a biologically-579 inspired robotic predator swimming at a varying speed and performing targeted attacks 580 581 elicits a strong antipredator behavioural response that erodes energy reserves and 582 compromises the body condition of mosquitofish. We propose that further efforts should 583 test whether biologically-inspired robots can effectively represent a novel, autonomous, and 584 effective solution to contrast the negative impact of invasive mosquitofish on freshwater 585 ecosystems worldwide [5-9].

586

587 AUTHOR CONTRIBUTIONS

GP and MP conceived the research question and supervised the research. GP designed the experiment. GP, VRS, and CS developed the experimental setup. MK, VRS, and CS conducted the experiments. GP and MK analysed the data and all authors discussed the results. GP and MK wrote a first draft of the Materials and Methods section. GP and MP wrote the manuscript. All the authors reviewed the final draft.

593

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598

599 DATA ACCESSIBILITY

All data can be found at <u>https://figshare.com/s/1e755aeba5b781f753a9</u>.

601

602 **ETHICS**

Experiments were performed in accordance with relevant guidelines and regulations and were approved by the University Animal Welfare Committee (UAWC) of New York University under the protocol number 13-1424. Notably, pilot tests on predator fish were approved through an animal care permit (G 0074/15) granted by the Landesamt für Gesundheit und Soziales Berlin (LAGeSo) and performed in Germany.

608

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- 612
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- 815

816 **TABLES**

817 Table 1 Analysis of variance with Satterthwaite's method from linear mixed models with

818 distance moved, freezing, speed variance, mean distance from replica, predator

819 inspection, thigmotaxis, and transfer entropy as dependent variables.

Fixed factors	Mean Square	df	F	Р
Distance moved (cm)				
К	2801434	1, 424	3.567	0.059
Mass	239961	1, 121	0.3055	0.581
Sex	2306162	1, 78	2.936	0.091
Week	82525615	1, 379	105.080	<0.001***
Condition	32234416	5, 367	41.044	<0.001***
Freezing (s)				
К	11392	1, 340	0.359	0.549
Mass	57781	1, 96	1.821	0.180
Sex	85910	1, 75	2.708	0.104
Week	3161017	1, 386	99.631	<0.001***
Condition	735543	5, 367	23.183	<0.001***
Speed variance (cm ² /s ²)				
К	3.376	1, 234	7.044	0.008**
Mass	0.047	1, 85	0.099	0.754
Sex	0.027	1, 74	0.056	0.814
Week	3.881	1, 389	8.099	0.005**
Condition	8.935	5 <i>,</i> 368	18.644	<0.001***
Distance from replica (cm)				
К	62.46	1, 186	3.784	0.053
Mass	0.390	1, 77	0.024	0.878
Sex	0.06	1, 73	0.003	0.954
Week	0.59	1, 315	0.036	0.850
Condition	343.970	4, 291	20.837	<0.001***
Predator inspection (counts)				
К	199.50	1, 288	3.025	0.083

Mass	66.50	1, 80	1.008	0.318
Sex	212.90	1, 74	3.227	0.076
Week	441.80	1, 314	6.698	0.010*
Condition	5539.30	4, 291	83.977	<0.001***
Thigmotaxis (s)				
К	59307	1, 404	2.944	0.087
Mass	23814	1, 107	1.182	0.279
Sex	104813	1, 75	5.202	0.025*
Week	232953	1, 380	11.562	<0.001***
Condition	291277	5, 365	14.457	<0.001***
Transfer entropy (bits)				
Condition	<0.001	2, 146	0.514	0.599
Direction	<0.001	1, 219	49.516	<0.001***
Condition*Direction	<0.001	2, 219	5.262	0.006**

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Fulton's *K*, body mass, sex, week, and condition are included in all models as fixed factors, except for transfer entropy in which condition, direction, and their interaction were included as fixed factors. The individual is included as a random effect (that is, random intercepts) in all models, while pair (fish-robot) is included as a second random effect in the transfer entropy model, to account for repeated measures. The significance was set at α <0.05, and significant results are indicated with * (<0.05), ** (<0.01), and *** (<0.001). Table 2 Results from general linear mixed models with distance moved, freezing, speed variance, mean distance from replica, predator inspection, thigmotaxis, and transfer entropy as dependent variables.

$V_{ m within}$	V_{among}	Repeatability	ΔΑΙϹ	χ^2_1	Р
785359	599257	0.433	118.770	120.769	<0.001***
31727	9038	0.222	33.978	35.977	<0.001***
0.479	0.035	0.069	2.322	4.322	0.038*
16.507	0.482	0.028	1.461	0.539	0.463
65.960	19.270	0.226	25.019	27.019	<0.001***
20148	11185	0.357	80.800	82.800	<0.001***
<0.001	<0.001	0.172	1.7	3.734	0.053
<0.001	<0.001	0.455	42.100	44.068	<0.001***
	V _{within} 785359 31727 0.479 16.507 65.960 20148 <0.001 <0.001	Vwithin Vamong 785359 599257 31727 9038 0.479 0.035 16.507 0.482 65.960 19.270 20148 11185 <0.001	Vwithin Vamong Repeatability 785359 599257 0.433 31727 9038 0.222 0.479 0.035 0.069 16.507 0.482 0.028 65.960 19.270 0.226 20148 11185 0.357 <0.001	V_{within} V_{among} RepeatabilityΔAIC7853595992570.433118.7703172790380.22233.9780.4790.0350.0692.32216.5070.4820.0281.46165.96019.2700.22625.01920148111850.35780.800<0.001	V_{within} V_{among} Repeatability Δ AIC χ_1^2 7853595992570.433118.770120.7693172790380.22233.97835.9770.4790.0350.0692.3224.32216.5070.4820.0281.4610.53965.96019.2700.22625.01927.01920148111850.35780.80082.800<0.001

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Fulton's K, body mass, sex, week, and condition are included in all models as fixed factors, 831 except for transfer entropy in which condition, direction, and their interaction were 832 included as fixed factors. Random intercepts are included for each individual in all models, 833 while random intercepts for each pair (fish-robot) are included for transfer entropy only, 834 835 which allowed variance decomposition. Within-individual variance (V_{within}), among-individual variance (V_{among}), and repeatability are shown for each behavioural trait. Test statistics (χ_1^2) 836 837 and significant levels of the random effects (that is, intercepts) were estimated using LRTs (P) and AICs between the full and the null model. Note that Δ AIC corresponds to the 838 difference in AIC between the null models minus the AIC from the full model. The 839 significance was set at α <0.05, and significant results are indicated with * (<0.05) and *** 840 841 (<0.001).

842 **FIGURE CAPTIONS**

Figure 1 Schematic for (A) the overview of experimental apparatus, (B) the 3D representation of the biologically-inspired predator replica, (C) the biologically-inspired predator replica used for experiments, and (D) a picture of a juvenile largemouth bass.

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Figure 2 Estimated marginal mean (EMMs) differences represent adjusted mean differences 847 (+ SE) in predator inspection (A) and thigmotaxis (B) across conditions once the contribution 848 of fixed effects included in the model (that is, Fulton's K, body mass, sex, week) is accounted 849 850 for, except sex that was excluded in EMMs for predator inspection to preserve positive 851 values in PM condition and favour the interpretation while not altering results. White 852 histograms correspond to control conditions (NP and PM), light grey histograms to openloop conditions (OL1 and OL2), and dark grey histograms to closed-loop conditions (CL1 and 853 854 CL2). NP condition is not shown for predator inspection (A) since fish were tested in absence 855 of the predator replica. Means not sharing a common superscript are significantly different. The significance was set at α <0.05. 856

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Figure 3 Transfer entropy between fish and robotic replicas. Transfer entropy from fish to robot are represented in panels A, C, and E and from robot to fish in panels B, D, and F with respect to conditions OL1 (first row), CL2 (second row), and CL1 (third row), respectively. Transfer entropy in both directions (fish-to-robot and robot-to-fish) for each of the three conditions is represented in panel G (+ SE). Means not sharing a common superscript are significantly different. The significance was set at α <0.05.

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Figure 4 Estimated marginal mean (EMMs) differences represent adjusted mean differences (+ SE) in Fulton's condition factor *K* across conditions once the contribution of fixed effects included in the model (that is, sex and week) is accounted for. The white histogram corresponds to controls (baseline, NP, and PM), the light grey histogram to replicas with low biomimicry (OL1, OL2, and CL2), and the dark grey histogram to replicas with high biomimicry (CL1). Means not sharing a common superscript are significantly different. The significance was set at α <0.05.

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- 874 FIGURES
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