1	Body condition and background noise alter female responses to uni- and
2	multimodal signals emitted by a male mimicking robot frog
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23 1. Mate choice in females is influenced by intrinsic and extrinsic factors, including 24 signal conspicuity, receiver body condition, and environmental properties. These 25 factors interact in complex ways to modulate the choice of mates. Multimodal signals 26 are more conspicuous than their unimodal components and therefore should elicit a stronger response. However, variations in female body condition and background 27 28 noise can modify their responsiveness to signals of varying conspicuity. 29 2. Males of the diurnal stream-dwelling frog Crossodactylus schmidti emit unimodal e 30 multimodal signals under variable noisy conditions, and the females vary greatly in 31 body condition. We tested hypotheses on how signal type (unimodal or multimodal) 32 interacts with female body condition and background noise to modify female 33 responses. In a field experiment using a male mimicking robot frog, females were 34 randomly exposed to acoustic-only (call, A), visual-only (toe flag, V), and multimodal 35 (call + toe flag, M) stimuli, while female body condition was estimated and data on 36 background noise was taken.

37 **3.** Females exhibited three types of response: emission of acoustic signals, emission of 38 visual signals, and movements toward the robot. All stimuli elicited responses, with a 39 higher percentage of females responding to M, an intermediate percentage to A, and a 40 lower percentage to V. Females in better physical condition exhibited a decreased 41 probability of acoustic response, emitted signals at a lower rate, and took more time to 42 emit signals or move towards the robot. With increasing noise, females exhibited 43 decreased probability of responding with a visual signal to both the visual and 44 multimodal stimuli, but when stimulated by the acoustic stimulus, females exhibited a 45 higher probability of visual response as the noise increased. Finally, females at noisier sites also emitted signals at a higher rate but took more time to respond with visual 46 47 signals and to move towards the robot. 48 4. The results suggest that the multimodal signal had the highest female 49

responsiveness. The negative effect of body condition and the positive effect of
background noise on the response occurred because better-conditioned females were
more selective in their choice, while noisy environments negatively interfered with
mate assessment. Our study highlights the complex and context-dependent nature of
female mate choice, influenced by signal conspicuity, female body condition, and noise
levels.

55 KEY WORDS: background noise, electromechanical model, female choosiness, female
56 quality, multimodal signaling, sexual selection.

57 INTRODUCTION

Mate choice is a pervasive pattern in sexually reproducing species and can be 58 59 described as a nonrandom allocation of reproductive resources from one sex to 60 individuals of the opposite sex (Edward, 2014). Although there is growing evidence that males also engage in mate choice (Bonduriansky, 2001; Edward and Chapman, 61 2011), this behavior is more intensively studied and understood in females (Andersson 62 1994; Rosenthal, 2017). Regardless of the sex, mate choice is the manifestation of the 63 mating preference, which is an innate tendency of individuals of one sex to accept 64 mating with individuals of the other sex bearing certain phenotypes (Jennions and 65 Petrie, 1997). However, mate choice is also influenced by intrinsic (condition-66 67 dependent) and extrinsic (context-dependent) factors, which affect which individuals are chosen as mates (Cotton et al., 2006; Dougherty, 2023). 68 Intrinsic factors capable of modulating mate choice include, for instance, age, 69 attractiveness, parasite load, and body condition (examples in Dougherty, 2023). The 70 71 influence of body condition (sensu Rowe and Houle, 1996) on mate choice can be 72 explained in terms of costs. Females in good body condition can afford to be choosy and spend significant time and energy searching and evaluating potential mates 73 (hereafter mate sampling). Conversely, females in poor body condition are less likely to 74 75 pay the costs of mate sampling and are expected to mate sooner than females in good 76 condition (Cotton et al., 2006). In fact, lower choosiness (also known as preference strength) in females in poor condition has been reported in some studies. For example, 77 78 an experimental study with black-field cricket, Teleogryllus commodus, showed that 79 females in poor condition have lower choosiness for high-quality male calls than

80 females in good condition (Hunt et al., 2005). Similarly, females of the canary Serinus

canaria that have their body condition experimentally reduced show lower choosiness 81 82 for high-quality male songs than females in good condition (Lerch et al., 2013).

Extrinsic factors affecting female mate choice include, for instance, male density, 83 84 male-male competition (Cotton et al., 2006) and environmental factors hindering 85 detection or discrimination of male sexual signals (e.g., Halfwerk et al., 2017; Cronin et al., 2019; Taylor et al., 2021; Wilgers et al., 2022). Background noise is an extrinsic factor 86 87 that can limit female detection and discrimination of acoustic signals. To overcome 88 this, males can use multimodal communication, such as a combination of acoustic and 89 visual signals, to increase their chances of being detected and chosen by females 90 (Rosenthal et al., 2004). Multimodal signals are considered more efficient than unimodal signals, as they are received through multiple sensory modalities (Partan 91 92 and Marler, 2005). They may evolve through efficacy-based selection, improving the 93 probability or intensity of receiver response in varying environmental conditions, 94 including noise (Hebets and Papaj, 2005). Studies show that females in many species 95 prefer multimodal over unimodal signals (Elias et al., 2006; Girard et al. 2015; Kozak and Uetz, 2019), but few have explored how background noise affects female mate 96 97 choice in response to uni- and multimodal signals (e.g., Reichert and Ronacher, 2014). 98 In general, the interaction between intrinsic and extrinsic factors and the variation in conspicuity of different signals emitted by a prospective mate are expected 99 100 to play a key role in modulating female mate choice. Consider a species in which males emit visual (v), acoustic (a), or multimodal (m) signals to attract females. In this 102 example, the signal *m* is the most conspicuous to females because it stimulates two 103 sensory modalities. Signal v is the least conspicuous because the receiver must be 104 oriented toward the signaler to perceive the signal. Finally, the conspicuity of signal *a*

101

105 is intermediate between m and v because it stimulates females but suffers from

background noise interference (Figure 1a). When we introduce inter-individual 106 107 variation in an intrinsic factor, such as female body condition, choosiness to different 108 signals can be modified. The differential costs of mate sampling for females in good 109 and poor condition should make the former choosier than the latter, as explained 110 above. However, regardless of females' body condition, their choosiness should be 111 higher for signal *m*, lower for *v*, and intermediate for *a* (Figure 1b). Similarly, mate 112 choice can be modified by natural variation in an extrinsic factor, such as background 113 noise. The conspicuity of the signal *m* and *a* is negatively affected by the background noise, decreasing females' response to them, regardless of their body condition. Given 114 115 that the multimodal signal also includes a visual component, the negative effect of 116 background noise on *m* should be lower than on signal *a*. Finally, because the signal *v* 117 is not affected by background noise, females' response to them should not be affected 118 (Figure 1c).

119 Here we used a male mimicking robot frog to test the influence of female body 120 condition and environmental background noise on female mate choice in Crossodactylus 121 schmidti, a diurnal frog that inhabits and reproduces in streams in the austral 122 Neotropics (see 'Study species' below). We tested the following hypotheses: (1) given 123 that signals from different sensory modalities vary in conspicuity (Endler 1992) and 124 multimodal signals tend to be more conspicuous than their unimodal components 125 (Bradbury and Vehrencamp, 1998; Partan and Marler, 2005), the female response 126 should be higher to the multimodal (more conspicuous signal), intermediate to the 127 acoustic, and lower to the visual signal (less conspicuous signal) (Figure 1a); (2) 128 because the response to a sexual signal may be influenced by intrinsic factors (Cotton 129 et al., 2006), the better the condition of the females, the choosier they should be, and the 130 sequence of female choosiness in response to different signals should be: multimodal >

131	acoustic > visual, regardless of their body condition (Figure 1b); (3) because the
132	conspicuity of the multimodal signal is usually higher than that of the acoustic signal
133	alone (e.g., Rosenthal et al., 2004; Mitoyen et al., 2019), the negative effect of
134	background noise on female response to the multimodal signal should be lower than to
135	the acoustic signal, but the response to the visual signal should not be affected by the
136	background noise (Figure 1c).
137	
138	MATERIAL AND METHODS
139	Study area
140	The study was carried out in two streams in Turvo State Park (27°14'34.08"S,
141	53°57'13.74"W, 376 m a.s.l.), located in the municipality of Derrubadas, state of Rio
142	Grande do Sul, southern Brazil. The park covers an area of nearly 17,500 ha and is one
143	of the largest remnants of semideciduous forest in southern Brazil. The local climate is
144	characterized as subtropical subhumid with a dry summer between December and
145	March. The mean annual rainfall is 1,787 mm, the mean annual temperature is 18.8 $^\circ\mathrm{C}$
146	and mean temperature of the coldest month is 13.3 °C (Maluf, 2000). In the area, several
147	previous studies on the ecology and behavior of Crossodactylus schmidti have been
148	conducted (e.g., Caldart et al., 2014, 2016a,b, 2022).
149	
150	Study species

151 The diurnal Neotropical stream-dwelling frog *C. schmidti* is a suitable study system to

152 test the hypotheses outlined here. Males compete for and defend territories containing

153 rocks (signaling sites) and underwater chambers (oviposition sites) emitting visual,

acoustic, and multimodal signals (Caldart et al., 2011, 2014, 2022). Males emit an

155 audiovisual display composed of the simultaneous emission of calls and toe flags 156 (Caldart et al., 2011, 2014), i.e., up-and-down movements performed with the toes, in 157 which toes are lifted slowly from the substrate, showing the contrasting coloration 158 between their ventral and dorsal parts (Hödl and Amézquita, 2001; Hartmann et al., 2005; see also Supplementary Videos in Appendix S1). The display is used in both in 159 agonistic interactions with males and in courtship interactions (Caldart et al., 2014). An 160 161 experimental study using a male mimicking robot frog has shown that toe flags 162 emitted in association with aggressive calls appear to provide contextual information 163 that modifies the receiver's response in territorial contests (Caldart et al., 2022). However, the role of toe flags associated with calls directed at females remains 164 unknown. 165

166 Sexually mature individuals occur year-round (Caldart et al., 2019), and males are found in call and visual signaling activity from sunrise to sunset (Caldart et al., 167 168 2016b). Advertising males may emit the advertisement call and the aggressive call 169 independently, with long intervals between the two calls, or may emit both calls 170 sequentially, i.e., aggressive call followed by advertisement call (Caldart et al., 2011). 171 Sequential aggressive-advertisement calls were described to other species of the family 172 Hylodidae (e.g., Nascimento et al., 2005) and possibly compose a multiple signal, with 173 the aggressive call serving to keep conspecific males away and the advertisement call 174 serving to attract females (Caldart et al., 2011). High background noise levels caused by 175 waterflow impair part of the multinote trilled advertisement call of males (Caldart et 176 al., 2016a), which probably hampers sexual communication between males and females. 177

178 Courtship involves a female evaluation of signals emitted by the male. These179 signals include aggressive-advertisement calls directed to nearby males, visual signals

180 including limb liftings and more frequently toe flags, and various multimodal signals (i.e., acoustic and visual) directed to the approaching female (Caldart et al., 2014). 181 182 When a female is receptive, she moves toward the male and signals back using a low-183 amplitude call or visual signals, such as limb lifting, toe flagging, body jerking, and 184 running-jumping display (Caldart et al., 2014). Before diving to the oviposition 185 chamber, the female who has accepted a courting male interacts tactilely with him via 186 an amplexus-like stimulus (Caldart et al., 2019; see also Supplementary Videos in 187 Appendix S1).

188

189 Stimuli preparation and experimental procedure

We developed an electromechanical robot frog that simulates a *C. schmidti* male and programmed the robot to emit calls and toe flags, combined or in isolation. Then, we exposed the robot to females in the field to induce interactions. A general overview of the electromechanical robot, as well as films of the robot frog in action and the receivers' responses to it, are presented in Caldart et al. (2022).

195 To test our hypotheses, we programmed the robot to emit three types of stimuli: 196 (1) acoustic (calls only), (2) visual (toe flags only), and (2) multimodal (calls and toe 197 flags). The three types of stimuli had the same temporal structure, composed of an 8-198 min stimulus phase. The stimulus phase of the multimodal signal was composed of a 199 train of 12 s of aggressive-advertisement call with a simultaneous toe flag in the middle 200 of each call, followed by 28 s of silent interval. This sequence was repeated 12 times 201 during the stimulus phase. In the case of the acoustic stimulus alone, the robot emitted 202 aggressive-advertisement calls without toe flags. The acoustic stimulus consisted of a 203 multiple call recorded from an average-sized male used in the original call description 204 (Caldart et al., 2011), with temporal features modified in Adobe Audition to match

population means and standardized at a sound pressure level of 70 dB (Caldart et al.,
2011, 2016; see also Appendix S2). The exact moments of signal emission during the
stimulus phase were the same for the three types of stimuli. Finally, in the case of the
visual stimulus alone, the robot emitted toe flags without calls. The duration of the toe
flag movement programmed in the robot was 0.4 s, which is equal to the average
duration of the signal as reported in Caldart et al. (2014) and used in a previous
experiment (Caldart et al., 2022).

212 We located and captured females in November and December 2017. Then we 213 registered the snout-vent length (SVL, precision 0.01 mm) and body mass (precision 0.1 214 g) of the females and marked them with a temporary cotton waist belt around the 215 pelvic region containing an individual color code. After the release of the females to 216 their respective sites, we tied a degradable flag containing the individual code of the 217 female in the vegetation right above the site where each individual was captured. 218 Because females are territorial, we could locate the same females during the 219 experiment to expose each of them to more than one type of stimulus in a repeated 220 measure experimental design. One day after the marking procedure, we searched for 221 marked females between 09:00 and 17:00 h. After finding a marked female, we 222 observed the female for 5 min to ensure that they were active and positioned the robot 223 at a viewing distance of 70 cm, at an angle of 30° relative to the focal female (as in 224 Caldart et al., 2022). We then waited for 2 min for acclimatization and, if the focal 225 female was not disturbed jumping in the water, we exposed her to one of the three 226 experimental stimuli. We randomized the sequence of presentation of stimuli to each 227 female to avoid the influence of the order of exposure of stimulus type on females' 228 responses.

We recorded the entire trial for each focal female with a digital camcorder (Sony 229 Handycam HDR-CX405). Immediately after the footage, we measured the background 230 231 noise level (dB) 30 cm above the female site with a sound level meter (Instrutherm DEC 232 500; C weighting curve: 20–12,500 Hz, dB range: 35–130 dB). After finishing a trial, we 233 either waited for at least 30 min (plus 2 min of acclimatization) before assigning 234 another experimental stimulus to the same focal female - continuing the trial only if 235 the female had stopped interacting with the robot – or moved the robot to another site 236 and repeated the procedure with another female. We recorded 26 females, of which 19 were exposed to the three types of stimulus, 6 were exposed to at least two types of 237 238 stimulus (i.e., acoustic and multimodal), and one was exposed to only one type of 239 stimulus (i.e., acoustic). Thus, a total of 19 females were exposed to visual, 26 to the 240 acoustic, and 25 to the multimodal stimulus.

Body condition was estimated as the residuals of an ordinary least square
regression between log₁₀-transformedlog₁₀-transformed body mass and log₁₀transformed SVL. The residual index has been extensively used to estimate condition
in amphibians (reviewed in Brodeur et al., 2020), including a previous study with *C*. *schmidti* (Caldart et al., 2022). Positive residuals indicate that individuals have better
body condition than individuals with negative residuals.

247

248 Statistical analyses

249 We investigated whether the stimulus type, female body condition, and background

250 noise (predictor variables) affected female responses to the robot using model selection

251 of generalized linear mixed models (GLMMs) or random-effect (frailty) Cox

252 Proportional Hazards models (CPHs), depending on the type of response variable. In

the GLMMs and CPHs, female identity was included as a random factor to control for

repeated exposure of the same individual to different experimental stimuli. Moreover,
the two continuous predictors (i.e., body condition and background noise) were
standardized and centered to zero to produce comparable effect sizes (Schielzeth,
2010).

258 The response variables used in the GLMMs were: (1) probability of emission of 259 acoustic response (binomial distribution of errors with a cloglog link function), (2) 260 probability of emission of visual response (binomial distribution of errors and logit 261 link-function), which included the limb lifting, toe flagging, body jerking, and running-262 jumping display (Caldart et al., 2014), and (3) emission rate of acoustic and visual 263 signals (Poisson distribution of errors and log link function). The response variables 264 used in the CHPs were: (4) latency to acoustic response, (5) latency for visual response, 265 and (6) latency for moving toward the robot. Because some females did not respond 266 during the stimulus phase, the latency data were right censored (1 = nonrespondent 267 females, 2 = respondent females).

These six response variables above are related to female responsiveness to sexual 268 269 signals emitted by the robot. Regarding the stimulus type, we predicted that females 270 would show high probability and emission rate of acoustic and visual signals, as well 271 as lower latency to respond to the multimodal stimulus, followed by the acoustic and 272 visual stimuli, respectively. Regarding body condition, we predicted that this trait 273 would be negatively related to the probability and emission rate of acoustic and visual signals and positively related to the latency of female response. Assuming that females 274 275 in poor body condition will invest less in mates searching, they should be more 276 responsive than females in good body condition to an average quality male, such as 277 our robot whose body size, as well as to the emission of acoustic and visual signals 278 corresponding to the average values of males in the population (see 'Stimuli

preparation and experimental procedure' above). Finally, with respect to background
noise, we predicted that this trait would be negatively related to the probability and
emission rate of acoustic and visual signals, and also to the latency of female response
when the robot is emitting multimodal and acoustic signals, but not visual signals.
Moreover, female responsiveness is expected to be more affected by background noise
when the robot is emitting acoustic signals when compared to multimodal signals
because the latter is more conspicuous than the former.

286 For each response variable, our model selection procedure followed the steps287 described below:

288 (1) Fit the global model (response variable ~ stimulus type * body condition *

289 background noise + (1 | ind); notation using R language.

290 (2) Check the goodness-of-fit of the global model, i.e., retrieve the model conditional

291 R². If the global model had a good fit ($R^2 > 0.5$), we proceeded with the analysis

because the best-fitted model would also have a good fit (Symonds and Moussalli,

293 2011).

(3) Model all possible combinations of simpler models based on global model terms to

explore the explanatory power of competing models. Twelve models (the global model

296 plus 11 simpler models including a null model), all of them with clear biological

297 meaning, were fitted for each response variable.

298 (4) Calculate the $\triangle AIC_c$ and the Akaike weight (w_i) of every model and classify the

299 models using their ΔAIC_c , which is a modified version of the AIC recommended for

300 small sample sizes (Symonds and Moussalli, 2011).

301 (5) Accept the model with the lowest ΔAIC_c value and the highest Akaike weight (w_i)

as the best fitted model (Symonds and Moussalli, 2011).

303 (6) In cases in which there was more than one model with $\Delta AIC_c < 2$, select as the best approximating model the one with the highest Akaike weight (w_i) . Moreover, we also 304 305 used an average model approach, according to which we built a model composed of all 306 the terms included in the models with $\Delta AIC_c < 2$. The coefficients of the average model 307 are the average of the estimated coefficients in each of these models, weighted by the 308 weight of evidence of each model (Burnham and Anderson 2002). The model average 309 approach aimed to check if the estimates of the best ranked model (i.e., the one with 310 the highest w_i) are qualitatively similar to the estimates of the average model. 311 (7) Make diagnostic plots and tests for the best ranked models ($\Delta AIC_c < 2$) because AIC 312 is affected by overdispersion in the data (Symonds and Moussalli, 2011). For the 313 GLMMs, we tested the model's goodness-of fit, the significance of the dispersion 314 parameter, and the presence of zero inflation (when applicable) using the package 315 DHARMa (Hartig, 2020). For CPHs, we checked the proportional hazards assumption 316 using the package survival (Therneau 2020). 317 (8) If diagnostic plots and tests were fine, present the 95% confidence set of best-ranked models, i.e., models with cumulative Akaike weight ≤ 0.95 , and the average model, 318 when applicable (Burnham and Anderson, 2002). 319 320 GLMMs were built using the package *glmmTMB* (Brooks et al., 2017), and the CPH models were fitted using the package survival (Therneau, 2020). Complete results 321 322 and diagnostics for all models are presented in Appendices S3-S5. All statistical analyses were performed in software R version 4.0.2 (R Core Team, 2020). 323

324

RESULTS

326 General patterns of female response to robot stimuli

327	Females exhibited three types of response to the robot: emission of acoustic signals
328	(hereafter acoustic response), emission of visual signals (hereafter visual response), and
329	movements toward the robot. The proportion of responsive females was higher to the
330	multimodal stimulus, lower to the visual stimulus, and intermediate to the acoustic
331	stimulus, regardless of the type of female response. When all types of female response
332	are pooled, the general pattern remains qualitatively the same (Figure 2).
333	
334	Effects of stimulus type, body condition, and background noise on the probability
335	of female acoustic response
336	The best-ranked model to explain the probability of female acoustic response included
336 337	The best-ranked model to explain the probability of female acoustic response included the additive effect of stimulus type and body condition. This model was the only one
337	the additive effect of stimulus type and body condition. This model was the only one
337 338	the additive effect of stimulus type and body condition. This model was the only one with $\Delta AIC_c < 2$ (Table 1, Figure 3a) and had a high goodness-of-fit (conditional R ² =
337 338 339	the additive effect of stimulus type and body condition. This model was the only one with $\Delta AIC_c < 2$ (Table 1, Figure 3a) and had a high goodness-of-fit (conditional R ² = 0.83). When compared to the visual stimulus, the acoustic stimulus induced an
337 338 339 340	the additive effect of stimulus type and body condition. This model was the only one with $\Delta AIC_c < 2$ (Table 1, Figure 3a) and had a high goodness-of-fit (conditional R ² = 0.83). When compared to the visual stimulus, the acoustic stimulus induced an intermediate probability of response, while the multimodal stimulus induced the
337 338 339 340 341	the additive effect of stimulus type and body condition. This model was the only one with $\Delta AIC_c < 2$ (Table 1, Figure 3a) and had a high goodness-of-fit (conditional R ² = 0.83). When compared to the visual stimulus, the acoustic stimulus induced an intermediate probability of response, while the multimodal stimulus induced the highest probability of response and the visual stimulus induced the lowest probability
 337 338 339 340 341 342 	the additive effect of stimulus type and body condition. This model was the only one with $\Delta AIC_c < 2$ (Table 1, Figure 3a) and had a high goodness-of-fit (conditional R ² = 0.83). When compared to the visual stimulus, the acoustic stimulus induced an intermediate probability of response, while the multimodal stimulus induced the highest probability of response and the visual stimulus induced the lowest probability of response. The probability of female acoustic response decreased with female body

Effects of stimulus type, body condition, and background noise on the probability of female visual response

The best-ranked model to explain the probability of female visual response included 348 349 the interaction between stimulus type and background noise. This model was the only 350 one with $\Delta AIC_c < 2$ (Table 1, Figure 3b) and had a high goodness-of-fit (conditional R² 351 = 0.62). The multimodal stimulus induced the highest probability of female visual 352 response, the visual stimulus induced the lowest probability of female visual response, 353 and the acoustic stimulus induced intermediate probability values. The effect of 354 background noise on the probability of female visual response depended on the of 355 stimulus type (Figure 3b). Specifically, the probability of female visual response to 356 visual and multimodal stimuli decreased with increasing background noise, whereas 357 the probability of female visual response to the acoustic stimulus increased with 358 background noise. Detailed results of the GLMM performed using model selection are presented in Appendix S3. 359

360

Effects of stimulus type, body condition, and background noise on the emission rate of signals by females

The model to explain the emission rate of signals by females included the additive effect of stimulus type, body condition, and background noise, but the other two models that included interaction terms also had $\Delta AIC_c < 2.0$ (Table 1). However, our interpretation is based on the best ranked model (conditional R² = 0.80) because it has the highest Akaike weight and the lowest number of parameters. Furthermore, the best ranked model's estimates are qualitatively similar to the estimates of the average model's estimates (Figure S1 in Appendix S4), indicating that the inclusion of additional parameters through model averaging does not increase the explanatorypower of the model.

The best-ranked model indicates that the three types of stimulus had a similar effect on the emission rate of signals. The emission rate of signals decreased with female body condition and increased with background noise. In both cases, the effect was stronger for the acoustic stimulus, weaker for the visual stimulus, and intermediate for the multimodal stimulus, although the confidence intervals of the estimates overlap (Figures 3c-d). Detailed results of all GLMMs performed using model selection are presented in Appendix S3.

379

Effects of stimulus type, body condition, and background noise on the latency to female responses

382 The best-ranked model to explain the latency to female acoustic response included only 383 body condition, a predictor also present in the second best-ranked model, which includes the additive effect of background noise (Table 2, Figure 4a). Our interpretation 384 of the results will be based on the best-ranked model because this model has an Akaike 385 386 weight more than two times higher than the second model (Table 2) and the confidence 387 interval of its parameter estimates does not overlap zero. Furthermore, in the average 388 model, the background noise confidence interval overlaps zero, indicating that this 389 variable does not add more explanatory power to the model (Table S1 in Appendix S4). 390 Body condition affected female response, so that females in better body condition took 391 longer to respond with acoustic signals after the onset of robot stimuli (Figure 4a).

The best-ranked model to explain the latency of the visual response included the additive effects of body condition and background noise, predictors that were also present in four of the five best ranked models (Table 2, Figure 4b). However, our 395 interpretation of the results will be based on the best-ranked model because this model 396 has an Akaike weight higher than the other models (Table 2) and the confidence 397 interval of the parameters' estimates does not overlap zero. Furthermore, contrary to 398 the best-ranked model, the confidence interval of all parameters in the average model 399 overlaps zero, indicating that the inclusion of additional parameters through model 400 averaging does not increase the explanatory power of the model (Table S1 in Appendix 401 S4). Females in better body condition and in sites with more background noise took 402 longer to respond with visual signals after the onset of robot stimuli (Figures 4b-c).

403 The best-ranked model to explain the latency to moving toward the robot 404 included the additive effects of stimulus type, body condition, and background noise 405 (Table 2). Among the four best ranked models, stimulus type appeared in all of them, 406 body condition in three of them, and background noise in two of them (Table 2). Our 407 interpretation of the results will be based on the best ranked model because this model 408 has an Akaike weight slightly higher than the other models (Table 2), and the 409 confidence interval of the parameters' estimates does not overlap zero. As in the 410 previous analyses, the confidence interval of body condition and background noise in 411 the average model overlaps zero, indicating that the inclusion of additional parameters 412 through model averaging does not increase the explanatory power of the model (Table 413 S1 in Appendix S4). Females in better body condition and in sites with more 414 background noise took longer to move toward the robot (Figure 4c-d). Finally, the 415 stimulus type affected the latency of female movement toward the robot, since for all 416 types of stimulus more than 50% of the females moved toward the robot (Figure 5c). 417 Females took longer to move toward the robot when the robot emitted the visual 418 stimulus (median latency = 364 s) but moved much faster toward the robot when it 419 emitted the multimodal stimulus (median latency = 50 s) or the acoustic stimulus

420 (median latency = 58 s). Detailed results of all CPH models performed using model
421 selection on latency data are presented in Appendix S5.

422

423 DISCUSSION

424 To test hypotheses on how body condition and background noise modify females' 425 responsiveness to signals of varying conspicuity, we innovate by presenting a male 426 mimicking robot frog to females of the diurnal stream-dwelling frog Crossodactylus schmidti. Our main results are: (a) females are more responsive to the multimodal 427 428 signal than to visual and acoustic signals, (b) females in good body condition are less 429 responsive than females in poor body condition, and (c) females at noisier sites exhibit 430 decreased probability of responding with a visual signal (except in response to the acoustic stimulus), emit signals at a higher rate, and take more time to signaling or 431 432 moving towards the robot. In what follows we explain how these results support 433 predictions about the effects of signal conspicuity and intrinsic and extrinsic factors on 434 female mate choice (Figure 1) and contribute to understanding how mate choice is shaped by the interplay between signal conspicuity, female body condition, and 435 436 background noise.

437 For any type of response we evaluated, females were more responsive to the multimodal stimulus, followed by the visual and acoustic stimuli, respectively. This 438 439 finding supports our hypothesis that signal conspicuity influences female response to sexual signals emitted by males and confirms the prediction that the sequence of signal 440 441 conspicuity is multimodal > acoustic > visual. Although there are several studies showing that females are more responsive to multimodal signals compared with 442 443 unimodal signals (e.g., Elias et al., 2006; Estramil et al., 2014; Girard et al. 2015), only a 444 few experimental studies provide quantitative support for this sequence among

visually and acoustically oriented species (e.g., Reichert and Höbel, 2015; Laird et al., 445 2016). Given that the content of a signal cannot be perceived and processed if it is not 446 447 properly detected (Wilgers et al., 2022), multimodal signals can contribute to more 448 informed decision-making during mate choice when compared with unimodal signals. 449 In fact, multimodal signals are believed to be more advantageous because they convey 450 more information per unit time, increasing opportunities for effective communication 451 (Rojas et al., 2018; Kabir and Thaker, 2021). Thus, we suggest that females of C. schmidti 452 are more responsive to the multimodal signal because it is more easily detected, and 453 also carries additional information about male signalers. Our suggestion implies content-based and efficacy-based selection on the evolution of multimodal signals 454 455 mediated by female response (Hebets and Papaj, 2005). We stress that we do not have 456 information on what specific information toe flags accompanying calls carries during 457 mating interactions in the study species. However, our results indicate that the 458 integration of multiple sensory modalities in male signals enhances females' response. 459 Females in better body condition were less responsive to male signals, exhibiting 460 lower probability of acoustic response, emitting signals at a lower rate, and taking 461 more time to emit signals or move towards the robot. Female acoustic response, 462 particularly, decreased with body condition in the expected sequence of signal 463 conspicuity (multimodal > acoustic > visual), further supporting the hypothesis that 464 signal conspicuity influence female-male communication during sexual interactions. 465 The consistent pattern across different types of female response indicates a prevalent 466 negative effect of body condition on female responsiveness (a proxy of choosiness, see 467 Edward, 2015). Our findings are qualitatively similar to previous empirical studies 468 with a wide variety of other taxa showing that choosiness is a condition-dependent 469 trait in females (e.g., Milinski and Bakker, 1992; Hunt et al., 2005; Cotton et al., 2006;

470 Wilgers and Hebets, 2012; Lerch et al., 2013; Judge et al., 2014; but see Dougherty,

2023). In the case of *C. schmidti*, females in good body condition can invest more time
and energy in mate sampling and mate evaluation, leading to more selective responses,
which ultimately increase their chances of finding a high quality male. By signaling
less frequently, females in better condition demonstrate higher selectivity for potential
mates, which possibly helps them avoid mating with lower-quality males and increase
their chances of securing a high-quality mate.

477 Females at noisier sites exhibited response patterns that suggest different 478 mechanisms of noise interference on mate choice. With increasing noise, females 479 exhibited a decreased probability of responding with a visual signal to both visual and 480 multimodal stimuli emitted by the robot frog. This pattern can be explained by cross-481 modal sensory interference effects (e.g., Stoffels et al., 1985), such as when a sensory 482 input in one modality inhibits perception or attention to stimuli in another modality 483 (e.g., perceptual suppression in Ide and Hidaka, 2013). This effect is known to humans, 484 in which acoustic noise degrades visual orientation (Hidaka and Ide, 2015). Cross-485 modal sensory interference has also been reported in Caribbean hermit crabs, in which 486 anthropogenic acoustic noise disrupts response to visual predatory cues (Chan et al., 487 2010). More recently, a laboratory experiment with an anuran species showed that 488 acoustic noise disrupts female responses to visual and audiovisual stimuli during mate 489 choice (Zhu et al., 2022). Our findings obtained under natural conditions suggest the 490 presence of cross-modal sensory interference in an anuran that inhabits streams where 491 background noise is intense (Caldart et al., 2016a). With increasing stream-generated 492 noise, females of C. schmidti appear to undergo sensory suppression, hindering their 493 response to visual and multimodal stimuli. Since both stimuli involve the visual 494 component of toe flagging, noise-induced suppression may impede females from

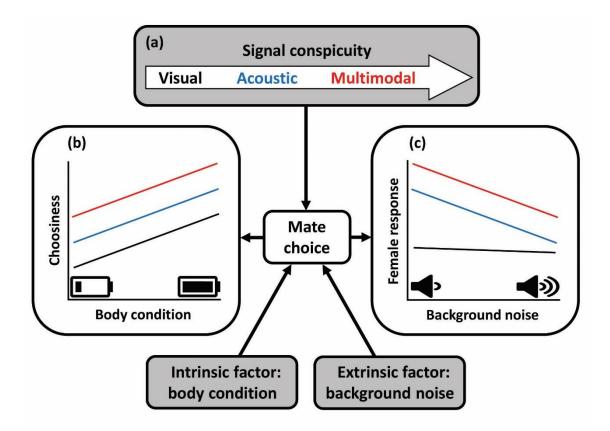
495 extracting important information from these signals, leading to a reduced496 responsiveness to them.

497 When exposed to the acoustic stimulus, females exhibited a higher probability of 498 visual response as the noise increased. A possible explanation for this seemingly 499 paradoxical result is that in anurans from lothic habitats, females have adapted their 500 sensory tuning to match the frequency range of male calls and filter out background 501 noise (Brumm and Slabbekoorn, 2005). However, we argue that females of C. schmidti 502 probably do not have such auditory adaptation because background noise negatively 503 affects their responses. Moreover, the background noise in the studied site partially 504 masks the multinote advertisement calls of males (Caldart et al., 2016a). Therefore, we 505 consider an alternative explanation for the higher responsiveness to acoustic stimuli 506 with increasing noise levels: females struggle to receive and interpret acoustic signals 507 in noisy conditions, and increase their visual responses and emit signals at a higher 508 rate to elicit more signaling from potential mates. Thus, heightened visual response of 509 females at noisier sites may be a strategy to overcome the challenges of receiving 510 acoustic signals in a noisy environment. The main implication of this strategy is that 511 females in noisy environments need to allocate more time and energy to signaling in 512 order to evaluate the quality of potential mates, which may also explain why females 513 in poor body condition are less choosy than females in good body condition.

Females at noisier sites also took more time to respond with visual signals and to move towards the robot, results that further indicate that the background noise influences female mate choice. When exposed to high levels of background noise, females may need more time to make an informed decision about male quality before responding because they are not receiving clear or sufficient information from the signaler (e.g., due to auditory masking) or because their sensory system is experiencing 520 reduced sensitivity (e.g., due to cross-modal sensory interference). Consequently, if 521 noise interference makes signals more difficult to detect via cross-modal sensory 522 interference or auditory masking, as our results suggest, females in noisier sites may 523 require more time to gather sufficient information from the signaler and decide to 524 respond or approach it. A recent study on the field cricket Gryllus bimaculatus found 525 that females exposed to anthropogenic noise took longer to approach and mate with 526 males, suggesting that noise interfered with their ability to make informed decisions 527 about mate quality (Bent et al., 2021). The delayed response was attributed to the need 528 for additional processing time in the presence of noise.

529 In conclusion, the results of our field experiment using a robot frog highlight the 530 importance of signal conspicuity in shaping female mate choice, with females being 531 more responsive to the multimodal stimulus compared to the unimodal stimuli. This 532 underscores the importance of integrating multiple sensory modalities into male 533 signals to improve detectability and elicit stronger female responses. Our results also 534 reveal the role of female body condition in influencing mate choice. The consistent 535 patterns observed in different females' behavioral responses indicate the strong 536 negative effect of body condition on female mate choice. A better body condition was 537 consistently associated with a lower probability of response, a lower signaling rate, and 538 a longer time to respond to the signals emitted by the robot frog. This finding supports 539 the notion that female mate choice in a condition dependent trait, so that females in 540 good body condition are choosier than females in poor body condition. Additionally, 541 we found that background noise has important effects on females' behavioral 542 responses. Females at noisier sites were less responsive overall, except in response to 543 the acoustic stimulus, emitting signals at a higher rate and requiring more time to 544 respond. This finding can be attributed to different mechanisms of noise interference

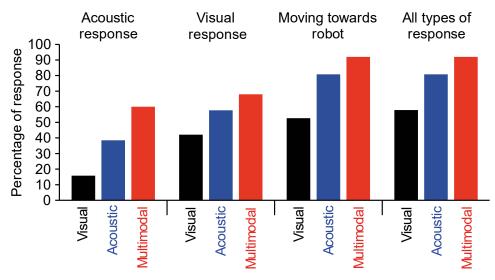
which deserve further investigation: 1) cross-modal sensory interference, as females at 545 noisier sites decreased response to stimulus presenting the visual stimulus, suggesting 546 547 that background noise hampers the perception and processing of visual signals; and 2) auditory masking, as females struggled to receive and interpret the acoustic signal in 548 549 noisy conditions and increased their response in order to elicit more signaling from 550 males. Finally, our study emphasizes that signal conspicuity, female body condition, 551 and noise levels shape female-male sexual interactions, revealing the complexity and 552 context-dependent nature of the mechanisms underlying female mate choice.



553

554 Figure 1

555 Hypotheses tested in this study: (a) given that signals from different sensory modalities 556 vary in conspicuity, females' response should be higher to the multimodal (more 557 conspicuous), lower to the visual signal (less conspicuous), and intermediate to the 558 acoustic; (b) because the response to a signal may be influenced by the receiver's body 559 condition, with high-quality females being choosier than low-quality females, the 560 response to all signals should decrease as female body condition increases, although 561 differences in the responses between signals are maintained by variation in signal 562 conspicuity; (c) because the response to an acoustic signal may be impaired by 563 background noise, females' response to the acoustic and multimodal signal should 564 increase as the background noise decreases, and the probability of female response should be higher to the multimodal signal across the noise gradient; response to the 565 566 visual signal is unaltered because this signal is not affect by background noise. 567



568

569 Figure 2

570 Percentage of females of the frog *Crossodactylus schmidti* that responded with acoustic

571 signaling, visual signaling, or moving toward the male mimicking robot frog after the

onset of visual (N = 19, black bars), acoustic (N = 26, blue bars), and multimodal (N = 19)

- 573 25, red bars) stimuli emitted by the robot.
- 574

575 Table 1

576 Results of the model selection of generalized linear-mixed models used to investigate

577 the effect of type of stimulus (acoustic, visual, and multimodal), body condition, and

578 background noise on the probability of acoustic and visual response, as well as on the

579 emission rate of signals by *Crossodactylus schmidti* females to a male mimicking robot

580 frog. For each response variable, the list of models shown represent the 95% confidence

set of best-ranked models, i.e., models with cumulative Akaike weight $(w_i) \le 0.95$.

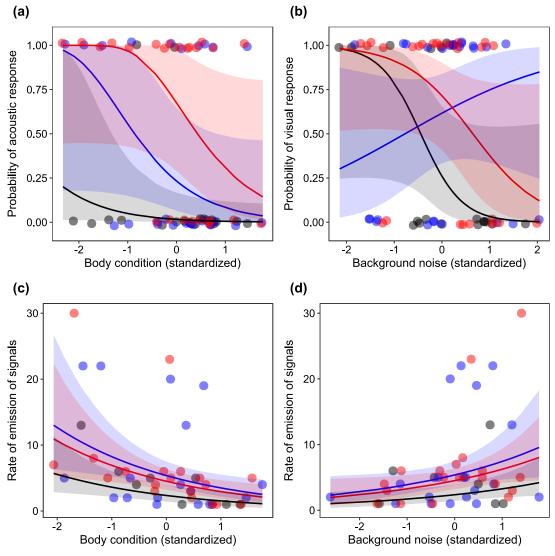
582

Response variable	Rank	Candidate model		AIC _c	ΔAIC_{c}	w_i
	1	stimulus type + condition		77.4	0.0	0.512
Probability of	2	stimulus type	4	79.6	2.3	0.164
acoustic response	3	stimulus type + condition + noise	6	79.7	2.3	0.161
	4	stimulus type * condition	7	80.5	3.2	0.106
	1	stimulus type * noise	7	94.4	0.0	0.564
	2	noise	3	98.5	4.1	0.072
	3	stimulus type * condition	7	98.7	4.3	0.067
Probability of	4	null	2	98.9	4.5	0.059
visual response	5	stimulus type + noise	5	99.3	4.9	0.048
visual response	6	stimulus type	4	99.7	5.3	0.041
	7	noise + condition	4	99.7	5.3	0.040
	8	condition	3	100.4	6.0	0.028
	9	stimulus type + condition + noise	6	100.6	6.2	0.025
	1	stimulus type + condition + noise	6	273.7	0.0	0.381
Emission rate of	2	stimulus type * condition * noise	13	275.0	1.3	0.195
signals	3	stimulus type * condition	7	275.3	1.6	0.170
SIGNAIS	4	stimulus type + condition	5	275.8	2.1	0.131
	5	stimulus type * noise	7	277.9	4.2	0.047

583 All models include female identity as random effect. Models are ranked by increasing 584 order of their ΔAIC_c values. The best models to fit the observed data ($\Delta AIC_c < 2.0$) are 585 indicated in boldface. k = number of parameters, ΔAIC_c = difference between the AIC_c

value of each model and the AIC_c value of the best-ranked model, and $w_i = AIC_c$

weight of each model. The symbol + indicates an additive effect between variables, andthe symbol * indicates an interaction between variables.



589 590 **Figure 3**

591 Effects of type of stimulus, female body condition, and background noise on the probability of response and emission rate of signals by Crossodactylus schmidti females 592 593 to a male mimicking robot frog. Stimulus-specific relationships between: (a) female 594 body condition and probability of acoustic response, (b) background noise and 595 probability of visual response, (c) body condition and emission rate of signals, and (d) 596 background noise and emission rate of signals. In all cases the stimulus emitted by a 597 male mimicking robot frog is represented in different colors: visual = black lines and 598 dots; acoustic = blue lines and dots; multimodal = red lines and dots. Tendency lines 599 shown here for each response variable are based on the best-ranked models (Table 1). 600

601 Table 2

602 Results of the model selection on random-effect (frailty) Cox Proportional Hazards

used to investigate the effect of type of stimulus (acoustic, visual, and multimodal),

body condition, and background noise on the latency to acoustic response, latency to

605 visual response, and latency to moving toward robot by *Crossodactylus schmidti* females

to a male mimicking robot frog. For each response variable, the list of models shown

607 represent the 95% confidence set of best-ranked models, i.e., models with cumulative 608 Akaike weight $(w_i) \le 0.95$.

609

Response variable Rank Candidate model		Candidate model	AICc	ΔAICc	k	wi
Latency to acoustic	1	condition	213.8	0.0	5.3	0.630
response	2	noise + condition	215.7	1.8	5.8	0.250
	1	noise + condition	292.5	0.0	8.6	0.292
Latency to visual	2	noise * condition	293.3	0.8	8.9	0.191
2	3	stimulus type + condition + noise	293.6	1.1	9	0.167
response	4	noise	293.8	1.3	9.5	0.153
	5	condition	294.3	1.9	10.1	0.115
	1	stimulus type + condition + noise	387.8	0.0	10.2	0.297
Latency to moving	2	stimulus type * condition	388.0	0.2	17.9	0.274
toward robot	3	stimulus type + noise	388.8	1.0	10.9	0.181
	4	stimulus type + condition	389.4	1.6	12	0.134

610 All models include female identity as random effect. Models are ranked by increasing

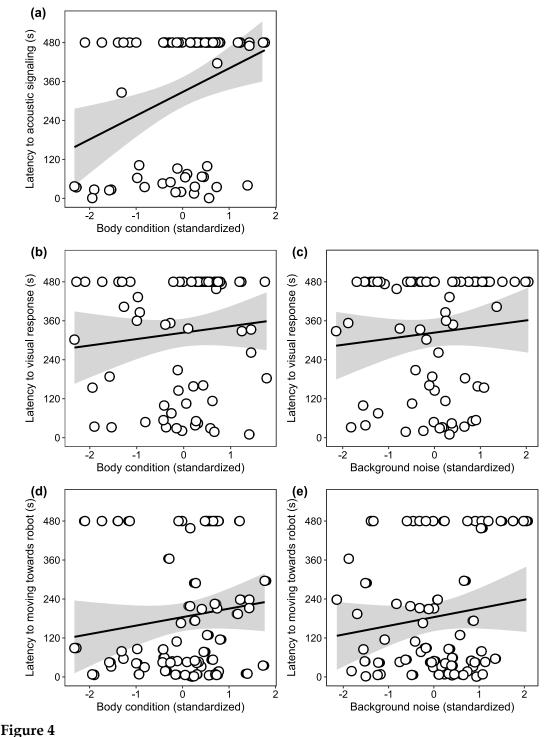
order of their ΔAIC_c values. The best models to fit the observed data ($\Delta AIC_c < 2.0$) are

612 indicated in boldface. k = number of parameters, ΔAIC_c = difference between the AIC_c

value of each model and the AIC_c value of the best-ranked model, and $w_i = AIC_c$

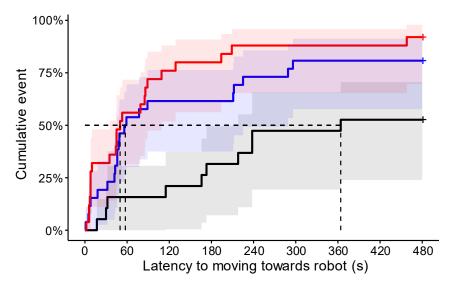
614 weight of each model. The symbol + indicates an additive effect between variables, and

615 the symbol * indicates an interaction between variables.



Fi

Effects of body condition on the latency of *Crossodactylus schmidti* females to respond to
the stimulus emitted by the male mimicking robot frog. Overall relationship between:
(a) female body condition and latency to acoustic response, (b) female body condition
and latency to visual response, (c) background noise and latency to acoustic response,
(d) female body condition and latency to moving towards robot, and (e) background
noise and latency to moving towards robot. Tendency lines shown here for each
response variable are based on the best-ranked models (Table 2).



625

626 Figure 5

627 Cumulative event probability curves for the latency of *Crossodactylus schmidti* females

628 to moving towards the robot frog in response to visual (black lines), acoustic (blue

629 lines), and multimodal (red lines) stimuli emitted by a male mimicking robot frog. A

630 vertical rise in the curves indicates that a response occurred at that time. Colored areas

631 indicate the 95% confidence interval for each curve. Dashed lines indicate the median

632 latency to respond to each experimental stimulus.

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637

638 CONFLICT OF INTEREST

639 The authors declare that they have no conflicts of interest.

640

641 AUTHOR'S CONTRIBUTIONS

642 VMC and GM conceived the study and designed methodology, VMC and MBS built

643 the robot, VMC collected and analyzed the data, VMC wrote the first version of the

644 manuscript. All authors revised the statistical analyses and multiple versions of the

645 manuscript.

646

647 DATA AVAILABILITY STATEMENT

648 Data will be available at Dryad Digital Repository upon manuscript acceptance.

649

650 ETHICAL NOTE

- This project was authorized by the license #21/2016 and authorization
- 4173/2016 for temporary marking of individuals, provided by the Divisão de

653 Unidades de Conservação da Secretaria do Ambiente e Desenvolvimento Sustentável

do Estado do Rio Grande do Sul. No individual was collected or held in captivity

655 during this study.

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