

1 **Sex change and effective population size: implications for population**
2 **genetic studies in marine fish**

3

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5

6 **Running title**

7 Sex-change and genetic estimates of N_e

8

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25

26 **Abstract**

27 Large variance in reproductive success is the primary factor that reduces effective population size (N_e) in
28 natural populations. In sequentially hermaphroditic ('sex-changing') fish, the sex ratio is typically skewed
29 and biased toward the 'first' sex, while reproductive success increases considerably after sex change.
30 Therefore, sex-changing fish populations are theoretically expected to have lower N_e than gonochorists
31 (separate sexes), assuming all other parameters are essentially equal. In this study, we estimate N_e from
32 genetic data collected from two ecologically similar species living along the eastern coast of South
33 Africa: one gonochoristic, the 'santer' sea bream *Cheimarius nufar*, and one protogynous (female-first)
34 sex-changer, the 'slinger' sea bream *Chrysolephus puniceus*. For both species, no evidence of genetic
35 structuring, nor significant variation in genetic diversity, were found in the study area. Estimates of
36 contemporary N_e were significantly lower in the protogynous species, but the same pattern was not
37 apparent over historical timescales. Overall, our results show that sequential hermaphroditism may affect
38 N_e differently over varying time frames, and that demographic signatures inferred from genetic markers
39 with different inheritance modes also need to be interpreted cautiously, in relation to sex-changing life-
40 histories.

41

42 **Introduction**

43 The amount of genetic diversity in a species reflects the effective population size (N_e), which in practical
44 terms informs on the number of breeders that contribute to the offspring, generation after generation; this
45 number, especially in some marine animals, has been estimated to be several orders of magnitude lower
46 than census size (N_c) (Hauser and Carvalho, 2008). Census size fluctuations and ecological perturbations
47 are known to reduce N_e , especially in fragmented populations. However, life history traits also play a
48 fundamental role in determining the effective size of a population (reviewed in Caballero, 1994). Lee et
49 al. (2011) suggest that delayed age-at-maturity and lowered juvenile survival reduce N_e/N_c . Recently, age-
50 at-maturity and adult lifespan were shown to explain half of the variance in N_e/N_c among 63 animal and
51 plant species (Waples et al., 2013). Thus, variation in key life history traits related to mating success and
52 survival rates, through their effect on individual lifetime reproductive success, appear to shape N_e
53 differences among populations.

54 In teleosts, older and larger females generally produce more and larger eggs (Chambers and
55 Leggett, 1996; Palumbi, 2004). Larger size also improves the mating capacity of males through
56 behaviours such as dominance and protection of territories (Warner, 1988). Hence, growth rate is also
57 likely a key factor determining reproductive success. However, additional reproductive traits in marine
58 fish might account for further components of lifetime variance in reproductive success. Sequential
59 hermaphrodites first mature as one sex and, after changing, reproduce as the opposite sex. Since younger
60 and smaller individuals of the first sex are generally more abundant than older and larger individuals of
61 the second sex, sequential hermaphrodite species typically present skewed sex ratios compared to
62 gonochoristic (separate sexes) species (Allsop and West, 2004). According to the size advantage model
63 (Ghiselin, 1969; Warner, 1975), reproductive success in sequential hermaphrodites tends to increase
64 considerably after sex change, with individuals of the ‘second sex’ expected to have a greater contribution
65 to the next generation. Therefore, the age at which individuals change sex – which has been shown to
66 fluctuate in response to environmental factors (Hamilton et al., 2007; Mariani et al., 2013) – will have a
67 significant impact on the lifetime reproductive success in sequential hermaphrodites. As a result,
68 intrinsically biased sex ratios (Wright, 1931) and variance in reproductive success are theoretically
69 expected to result in lower N_e in sequentially hermaphroditic species, compared to gonochoristic ones,
70 assuming that other characteristics are somewhat equal. This is inevitably complicated by the flexible

71 nature of age/size-at-sex change in natural populations (see Avise and Mank, 2008; Mariani et al., 2013;
72 Ross, 1990, for discussion).

73 In the present study, by directly examining empirical data, we investigate whether sex-changing
74 life history may indeed determine a reduction in N_e as a result of increased lifetime variance of
75 reproductive success and skewed sex ratio. We compared genetic data in two closely related and
76 sympatric species (family: Sparidae) with largely comparable habitat, ecology, abundance, behaviour and
77 life-history traits, with the exception of their reproductive modes: one species being protogynous (the
78 slinger sea bream, *Chrysoblephus puniceus*, which first matures as female, later turning to male) and the
79 other being gonochoristic (santer sea bream, *Cheimereus nufar*, maturing either as male or female)
80 (Garratt, 1985a, b, 1986, 1993). *Chrysoblephus puniceus* is endemic and restricted to the south-east coast
81 of southern Africa, while *C. nufar* is distributed over a wider area of the Western Indian Ocean. Both
82 species are targeted by the same local commercial and recreational line fisheries, and together represent a
83 large proportion of catches from this region (Mann and Fennessy, 2013a, b). Both species are
84 opportunistic predators found in shoals around coastal reefs feeding on crustaceans, mollusks and small
85 fish (Garratt, 1986). Sex-ratios were found to fluctuate greatly in *C. puniceus*, both spatially, between
86 southern Mozambique and the KwaZulu-Natal region (Garratt, 1985a), and temporally (Mariani et al.,
87 2013), with changes to some extent influenced by the degree of fishing pressure. In contrast, even sex
88 ratios were found in *C. nufar* (Garratt, 1985a). Results obtained using nuclear and mitochondrial
89 molecular markers are interpreted as a function of inheritance mode and reproductive strategy. The
90 findings enhance our understanding of the role of life-history in population genetics and may have
91 implications for the management of exploited populations.

92

93 **Materials and methods**

94 *Sampling*

95 *Chrysoblephus puniceus* and *Cheimereus nufar* specimens were collected from commercial ski-boat line
96 fishermen between May and July 2007 at three locations along the KwaZulu-Natal coast: Port Edward
97 (PE), Park Rynie (PR) and Richards Bay (RB) (Fig. 1). A total of 138 *C. puniceus* (122 females and 16
98 males) and 139 *C. nufar* (69 females and 67 males) were collected.

99 Fork length and weight were measured and sex was assessed by macroscopic gonad identification. We
100 used previously published length-age relationships to derive age from fork length for both *C. puniceus*
101 (Garratt et al., 1993) and *C. nufar* (Druzhinin, 1975; Coetzee and Baird, 1981). Fin-clips were taken from
102 the pectoral fins and stored in absolute ethanol for later DNA extraction.

103

104 *Molecular analyses*

105 DNA was extracted using a modified Phenol-Chloroform protocol (Sambrook and Russell, 2001). DNA
106 concentration and quality were estimated on a NanoDrop™ ND-1000 spectrophotometer. Samples of
107 both species were screened at 11 microsatellite loci, some specifically developed for this study (Chopelet
108 et al., 2009a). Of the eleven microsatellites designed for *C. puniceus*, five cross-amplified in *C. nufar*. Six
109 supplementary microsatellites were specifically developed for *C. nufar*, using the same protocol as in
110 Chopelet et al. (2009a) (see Table 1 for details). Microsatellites were amplified using fluorescence-
111 labelled forward primers (Applied Biosystems, Waltham, Massachusetts) and 2X Multiplex PCR Master
112 Mix (QIAGEN, Hilden, Germany) in a final volume of 10µl. Depending on size and dye, fragments were
113 amplified into two multiplexed reactions for *C. puniceus* (Chopelet et al., 2009a), while another
114 microsatellite (SL3) was amplified separately. For *C. nufar*, one reaction contained SL25 and SA2, and
115 the other included the nine remaining loci (Table 1). All amplifications were carried out using the same
116 conditions. An initial step of 15 min at 95°C was followed by 30 cycles of 45s at 94°C, 45s at 60°C, and
117 45s at 72°C and a final extension step at 72°C for 45 min. PCR products were sized on an ABI 3130xl
118 alongside a GS600 ladder. Genemapper v 4 (Applied Biosystems) was used for allele scoring.

119 Universal primers Hsp1 and Lsp1 were also used to amplify the first hypervariable region of the
120 mitochondrial DNA control region (Ostellari et al., 1996). Each reaction was carried out using 300 ng of
121 genomic DNA in Ready Mix (Applied Biosystems) in a final volume of 25 µl. PCR cycles were as
122 follows: a) 95°C (5min), b) 30 cycles at 95°C (50sec), 56°C 1min and 72°C (2min), and c) a final 10 min
123 elongation at 72°C (10 min). Amplified products were purified with exonuclease I and shrimp alkaline
124 phosphatase (Hanke & Wink 1994) and sequenced at GATC-Biotech (Konstanz, Germany).

125

126 *Statistical analyses*

127 *Genetic diversity*

128 Genetic analyses were first performed to detect patterns of spatial structure and estimate diversity within
129 and between locations. For microsatellite data, frequencies of null alleles were estimated using FREENA
130 (Chapuis and Estoup, 2007). For each location sample, number of alleles (N_a), observed (H_o) and
131 expected (H_e) heterozygosity and inbreeding coefficient (F_{IS}) were assessed using GENEPOP ON THE WEB
132 (<http://genepop.curtin.edu.au/>; Raymond and Rousset, 1995; Rousset, 2008). Marker neutrality was tested
133 in LOSITAN (Antao et al., 2008). To correct for variance in the sample size among populations, we further
134 estimated the allelic richness (A_R) based on the minimum sample size (see Table 1) using FSTAT 2.9.3
135 (Goudet, 1995) and the number of private alleles using the rarefaction method (Kalinowski, 2005)
136 implemented in ADZE 1.0 (Szpiech et al., 2008). Population differentiation was estimated using the θ
137 estimator of F_{ST} (Weir and Cockerham, 1984) and relative confidence intervals using 10,000 permutations
138 on the individuals in GENETIX v.4.05.2 (Belkhir et al., 1996-2004). Although the mutation rate was in our
139 instance likely to be orders of magnitude smaller than migration rates, the corrected F'_{ST} (Hedrick, 2005)
140 was calculated, and Jost's D_{est} (Jost, 2008) estimated using SMOGD (Crawford, 2010), and reported as
141 additional information on genetic substructure. POWSIM was employed to evaluate the power of the
142 dataset to detect genetic differentiation (Ryman et al., 2006). Five-hundred replications and different N_e/t
143 ratios (500/0; 2000/2; 1000/5; 1000/10; 500/10) were used to obtain the expected F_{ST} according to this
144 equation: $F_{ST} = 1 - (1 - 1/2N_e)^t$, with t being the number of generations of isolation (Ryman et al., 2006).
145 Bayesian assignment was performed in STRUCTURE 2.3 (Pritchard et al., 2000; Falush et al., 2003, 2007)
146 to infer the most likely number of genetic clusters (K) present in the datasets using the admixture model,
147 and 500,000 iterations, after 50,000 burn-in. The number of clusters was calculated by averaging the
148 mean posterior probability of the data $L(K)$ over 10 independent runs.
149 Nucleotide and haplotype diversities were estimated from mitochondrial DNA sequences using DNASP
150 v4.5 (Rozas et al., 2003). Median-Joining networks were constructed for both species using POPART
151 (<http://popart.otago.ac.nz>).

152

153 *Estimating effective population size*

154 We used the linkage disequilibrium (LD) method implemented in LDNe (Waples and Do, 2008) to
155 estimate contemporary N_e for each location from the microsatellite data, both including and excluding the
156 markers that were not under HWE according to the exact test performed in Genepop. The LD method is

157 based on the theoretical relationship (Hill, 1981) between a measure of LD (r^2 = squared correlation of
158 alleles at pairs of unlinked gene loci), sample size (N), and N_e . LDNe implements a modification of Hill's
159 method that accounts for bias from ignoring second order terms in N and N_e . LDNe allows one to screen
160 out rare alleles, which tend to upwardly bias N_e estimates, by selecting a minimum allowable allele
161 frequency (P_{crit}). We focused especially on $P_{\text{crit}} = 0.02$, which (given the minimum sample sizes of N =
162 39-45; Table 2) ensured that any alleles that occurred in a single copy were not used (Waples and Do,
163 2010). Another estimate of effective population size was obtained using the Approximate Bayesian
164 Computation method implemented in DIYABC 2.01 (Cornuet et al., 2014). Calculations were performed
165 for each species, pooling samples from all locations in order to reflect the lack of genetic substructure
166 detected in our data. Three simple scenarios were simulated. Each one represented one single population
167 whose N_e remained constant (scenario 1), one where N_e increased after a time t_1 (scenario 2), and the third
168 where N_e decreased after t_1 . Priors were as follows: effective population size was between 10 and 10^6 , and
169 t_1 between 10 and 10^4 generations,

170 A longer-term view of effective population size was also obtained through estimates of historical
171 female N_e from mtDNA data. We first used the Watterson estimator of the mutation parameter theta (θ)
172 obtained from the number of polymorphic sites (S) (Watterson, 1975). In DNASP v4.5, θ is defined as
173 $2N_e\mu$ for mitochondrial DNA, where N_e is the effective population size and μ is the mutation rate per
174 DNA sequence per generation (Tajima 1996). We estimated the female effective population size (N_{ef})
175 from the haplotype mutation rate and generation time (T) according to this equation:

$$N_{ef} = \frac{\theta}{2\mu T}$$

176
177 We assumed a widely accepted rate $\mu=11\%$ per site per million year for the Sparid mtDNA
178 control region (Bargelloni et al., 2003; Sala-Bozano et al., 2009; Coscia et al., 2012), equal to 0.055
179 substitutions/site/million years. The age at maximum egg production was estimated with $L_{\text{inf}} = 47$ cm for
180 *C. puniceus* and $L_{\text{inf}} = 75$ cm for *C. nufar* (where L_{inf} , a parameter of the von Bertalanffy growth
181 equation, is defined as the length that an individual would reach if it grew indefinitely). This, according to
182 Beverton (1992), can be used as an approximation of generation time ($T = 5$ for *C. puniceus* and 7 for *C.*
183 *nufar*). Therefore, to account for life-history plasticity, we estimated N_e in both species with generation
184 time encompassing these values: $T = 3, 5$ and 8.

185 Furthermore, we applied the Bayesian Skyline Plot (BSP) approach implemented in BEAST v
186 1.7 (Drummond et al., 2012) to estimate trends in past effective population size. Firstly, jModelTest 0.1.1
187 (Posada, 2008; Guindon and Gascuel, 2003) was used to select the best model of substitution for each
188 dataset via the AIC (Akaike Information Criterion): GTR (Generalised Time Reversible described in
189 Tavaré (1986)) was selected for *C. nufar* and HKY (Hasegawa, Kishino and Yano 1985) for *C. puniceus*.
190 To avoid convergence issues, several independent runs (each 10^6 generations and 10% burn-in) were used
191 for each species until each effective sample size value (ESS) reached ~200 as per the user's manual.

192

193 **Results**

194 *Species and population characteristics*

195 Males (375-1404 g) were larger than females (257-952 g) in *C. puniceus*, while male (420-1436 g)
196 and female sizes (440-2820 g) overlapped in *C. nufar* (Figures S1 and S2). Individuals from the most
197 northern location, Richards Bay, had slightly larger sizes in both species (Fig. S2). Fifteen male *C.*
198 *puniceus* were found in Richards Bay and one male in Park Rynie, while only females were collected in
199 Port Edward (southernmost location). A recent study has shown that the likely age-at-sex-change for *C.*
200 *puniceus* is around 301mm, significantly lower than three decades ago (Mariani et al., 2013), and
201 potentially decreased as a result of increased fishing pressure.

202

203 *Genetic variation*

204 No evidence of null alleles was detected within the dataset. At least one marker in each of the two
205 species deviated from equilibrium expectations across all locations with strongly positive associated F_{IS}
206 values (SA3 in *C. nufar*, and SL35 in *C. puniceus*, Table S3). Since LOSITAN did not detect any signature
207 of selection (data not shown) and no significant F_{IS} was recorded for any species at each location (Table
208 2), all markers were retained in the subsequent analyses.

209 Expected and observed heterozygosities were 0.83 in all locations for *C. puniceus*, and varied
210 between 0.72 (PE and RB) and 0.74 (PR) for *C. nufar*. Allelic richness and the number of alleles were
211 higher in *C. puniceus* (overall $N_A=18.4$ and $A_R=18.1$) than *C. nufar* (overall $N_A=12$ and $A_R=12.1$) (Table
212 2). No significant F_{ST} values were found within either species (Table S1; *C. puniceus* overall $F_{ST} =$
213 0.0011, $p=0.19$; *C. nufar* overall $F_{ST} = 0.0004$, $p=0.37$). No significant genetic differentiation was found

214 between the three samples, for each species, irrespective of the method employed (overall D_{est} was 0.0036
215 for *C. puniceus* and 0.0010 for *C. nufar*) or corrections used (Hedrick's corrected G'_{ST} was 0.0047 for *C.*
216 *puniceus* and 0.0019 for *C. nufar*). Hence, as expected, STRUCTURE detected one genetic cluster in each
217 species ($K=1$, Fig S3). According to the power estimations implemented in POWSIM, the probability that
218 our datasets can detect low genetic differentiations up to an F_{ST} of 0.005 is 100% (Fig. S4), with high
219 probabilities (70-80%) also for values around 0.0025. This suggests a lack of genetic structuring among
220 the three locations sampled at microsatellites (see also Table S1).

221 A total of 75 individual *C. puniceus* and 77 *C. nufar* were sequenced (179 bp and 223 bp long
222 fragments of the control region respectively; Genbank Accession Nos.: XXXX–XXXX). Overall,
223 nucleotide diversity π was 0.03 for *C. puniceus* and 0.04 for *C. nufar*, while haplotype diversity H_d was
224 0.996 for *C. puniceus*, with 65 haplotypes, and 0.984 for *C. nufar*, with 55 haplotypes. Within species,
225 both π and H_d did not vary, with the former being 0.03 and 0.04 for *C. puniceus* and *C. nufar* respectively,
226 across all sampling locations (Table 2). H_d ranged between 1 (Richards Bay) and 0.996 (Port Edward) for
227 *C. puniceus*, and between 0.994 (Richards Bay) and 0.969 (Park Rynie) for *C. nufar*. No significant Φ_{ST}
228 was detected for any of the two species between any location (Table S1), and visual inspection of median
229 joining networks also showed a lack of geographical structure (Figures S5 and S6).

230

231 *Effective population size*

232 N_e estimates using LDNe tended to increase with extreme values of allelic frequencies (P_{crit}) (Fig.
233 2). However, *C. puniceus* consistently had much smaller N_e and smaller variances than *C. nufar*,
234 irrespective of the allelic frequencies included (Fig. 2). Estimates of effective population size (N_e^{\wedge}) at P_{crit}
235 = 0.02 are reported in Table 2. In *C. nufar*, negative N_e were interpreted as infinite (Waples and Do,
236 2010); the lower boundaries of the confidence intervals for *C. nufar* did not overlap with the highest
237 boundaries of the *C. puniceus* estimates, using all loci, and only marginally overlapped when we excluded
238 the locus not in HWE (Table 2). The pattern was found to be robust and consistent, whether samples were
239 pooled or treated as separate geographical collections. Historical N_e was estimated from microsatellites
240 using DIYABC. Of the three scenarios simulated, the second one was by far the most likely for *C.*
241 *puniceus* (N_e increased in time), while scenarios 1 and 2 were equally likely for *C. nufar*. For consistency
242 with *C. puniceus* data – and in line with population expansion results from mtDNA data (see below) – we

243 used scenario 2 for *C. nufar* too. Estimates of effective population size (t_0) were slightly higher for *C.*
244 *puniceus* than *C. nufar*, but the confidence limits were highly overlapped (Table 2).

245 Historical female population size inferred from mtDNA offered further insight into demographic
246 aspects, but the obtained estimates proved to be of different orders of magnitude depending on the method
247 used. The Watterson method returned estimates in the order of the tens of thousands (Fig 3), slightly
248 larger in *C. nufar* than in *C. puniceus*, all else being equal; however, considering the longer generation
249 time in the latter, estimates for the protogynous *C. puniceus* are slightly greater than for *C. nufar* (Fig 3).
250 The Bayesian Skyline method (Fig 4) also indicates larger effective size in *C. puniceus*, although the
251 confidence limits largely overlap.

252

253 **Discussion**

254 No significant genetic divergence was detected among populations of either species, using either
255 microsatellites or the variable mitochondrial control region, suggesting that the three sampled populations
256 for each species do not belong to independent evolutionarily significant units (ESUs). This lack of genetic
257 differentiation is also mirrored by similar levels of genetic diversity among locations. Similar findings for
258 *C. puniceus* were recently reported by Duncan et al. (2015) who reported a single panmictic stock. The
259 life histories and behaviour of both species are consistent with the lack of genetic differentiation within
260 this area: they live in large shoals in the vicinity of rocky and coral reefs and are capable of migration
261 across tens of kilometres (Garratt, 1984; Griffiths and Wilke, 2002). Based on studies of histological
262 changes in the ovaries, it appears that spawning in *C. puniceus* occurs in large aggregations off the coasts
263 of southern Mozambique and Northern KwaZulu-Natal (mostly to the north of Durban) and decreases to
264 the south of Park Rynie (Garratt, 1985). Most *C. puniceus* males sampled in this study were from
265 Richards Bay, in accordance with this pattern. The majority of eggs are fertilized and scattered above
266 coral and rocky reefs from northern locations, and the developing larvae drift southward influenced by the
267 strong Agulhas Current in this region (Garratt, 1985). Juveniles settling out to the south of Park Rynie,
268 where no spawning takes place, are believed to move northward later in life to reproduce (Maggs et al.,
269 2013). Accordingly, average sizes were found to be higher in the northernmost location of Richard's Bay.
270 Such southward larval dispersal followed by northward adult migration could result in a constant
271 redistribution of alleles across the area, ultimately ensuring population cohesion.

272 *Cheimarius nufar* has an extended distribution along the African east coast. The spawning area
273 stretches from the Eastern Cape to the Gulf of Aden and individuals in spawning condition are found
274 regularly in both Eastern Cape and KwaZulu-Natal waters (Coetzee, 1983; Garratt, 1985). Patterns of
275 adult migration and settlement of juveniles are less well known for *C. nufar* (Griffiths and Wilke, 2002),
276 but it is likely that other separate spawning aggregations exist in more northern, unsampled areas of the
277 Western Indian Ocean.

278

279 This study examined population genetic inference of N_e in the context of sex-change in marine
280 fish. Other sources of N_e variation were minimized by comparing two species with very similar
281 taxonomical and biological characteristics and by obtaining replicate samples from the same biological
282 units and dwelling in the same locations. No estimates of census size exist for these species, but these are
283 also likely to be similar in the study area: *C. puniceus* and *C. nufar* are known to be the two most
284 abundant commercial line-fishery targets in the South-African east coast, with annual landings around
285 300,000 and 100,000 individuals, respectively, but with the fishery that targets a habitats slightly more
286 suited to the former (Winker et al., 2012).

287 All methods to calculate effective population size will assume discrete generations; therefore,
288 underestimation of N_e may occur here, but will likely affect both species similarly (Waples et al., 2014).
289 Predictions from population genetics theory led to the expectation that a sex changing species would
290 present smaller N_e than a gonochoristic one for the following reasons: a) the sex ratio of the sex-changing
291 species is biased toward the 'first sex', which is known to reduce N_e (Wright, 1931; Chopelet et al.,
292 2009b); and b) the dynamic balance of age-at-sex change, which affects individual variance in
293 reproductive success (V_k). Using the LD method, we find agreement with these expectations, with
294 substantially lower effective sizes in the protogynous *C. puniceus*; while N_e estimates in *C. nufar* were
295 one order or magnitude larger, with infinite upper confidence boundaries, indicating that more precise
296 estimates in this species would ideally require greater sample sizes (Palstra and Ruzzante, 2008).
297 Obviously, the N_e/N_c ratios of these species may be different, but unpublished surveys indicate that local
298 abundances are in the same order of magnitude (Winker et al., 2012). Protogynous populations can still
299 therefore sustain large numbers through a relatively small number of breeders; over short time scales, a
300 population may be dominated by a relatively small number of strong, large breeding males, belonging to

301 two or three year classes, and able to pass their gene combinations to most of the new cohorts. On the
302 other hand, ABC computation provides no evidence that N_e differs significantly between breeding
303 strategies. Point estimates and confidence limits are very similar in the two species, suggesting that over
304 evolutionary time scales, some factors are at play in reducing the expected high variance in individual
305 reproductive success of sex-changers. One such scenario could be the inherently ‘bet-hedging’ nature of a
306 sex-changing life history, whereby successful genotypes with optimal timing of first maturation and sex-
307 change are able to maximize reproductive output as both females and males, hence reducing long-term
308 stochasticity.

309 Effective population size estimates based on mtDNA do not reveal substantial differences between
310 *C. nufar* and *C. puniceus*, except for the magnitude of the point estimates (i.e. millions with BEAST and
311 tens of thousands with the Watterson method). Although the Bayesian Skyline Plot shows overlapping
312 confidence limits, N_e appears larger in *C. puniceus*; with the Watterson method – once taken into account
313 that the generation time for *C. puniceus* is lower (Garratt, 1985, Mann et al., 2014) – there also seems to
314 be a trend towards greater N_e in *C. puniceus*. Interestingly, in a protogynous species like *C. puniceus*,
315 virtually every individual has the potential to reproduce as a female, with only a fraction becoming males.
316 This is likely to maintain a large mtDNA diversity over evolutionary timescales. In gonochoristic species
317 like *C. nufar*, only about half of individuals (the females) from every new cohort will transmit their
318 mtDNA, which may explain the patterns detected here using maternally inherited loci. Based on this,
319 protogynous species can be expected to have mtDNA diversity similar or higher, while protandrous
320 species would likely be lower, than that estimated for gonochoristic species.

321 This is potentially an important fact to be considered in conservation biology, and presumably
322 similar (yet opposite) considerations should be applied to protandrous species (where only a smaller
323 proportion of individuals will reproduce as females). This hypothesis can potentially be tested already
324 using the available literature. However, expectations will always require adjustment according to the
325 degree of plasticity of sex-changing patterns; for instance, a recent stock assessment of the *C. puniceus*
326 population in South Africa has revealed strong evidence of recovery since the introduction of a cut in
327 commercial fishing effort in 2003-2006 (Winker et al., 2012). This may have been facilitated by earlier
328 age-at-sex-change inferred by Mariani et al. (2013).

329 Much remains to be understood about the interaction among life history traits in determining
330 lifetime variance in reproductive success and, by reflection, the effective size of a population, especially
331 when the added complexities of sex-change, its timing and its extent are added to the picture. Beyond
332 fecundity, longevity and age at maturation, sex change is a particularly labile trait: individuals of the first
333 sex might sometimes be larger than individuals of the second sex, it is not always the largest individuals
334 of a group that change sex (Munoz and Warner, 2003), and some individuals in the population might not
335 change sex at all (Mann and Buxton, 1998). In the case of *C. puniceus* it is likely that the large females
336 that do not change sex make a significant contribution to future generations (Berkeley et al., 2004;
337 Palumbi, 2004). Furthermore, the social status of individuals within the group will affect the timing of
338 sex change, as will behavioural and/or biochemical (pheromones) interactions (Munday et al., 2006). Our
339 initial empirical analyses show that theoretical predictions may be supported over contemporary time
340 scales, but other mechanisms may play bigger roles over evolutionary periods.

341 In a population, it is generally possible to distinguish different groups such as individuals of the
342 same age or with the same sex. The reproductive success (the average number of gametes transmitted to
343 the next generations) of these groups can differ significantly (males can produce and transmit more
344 gametes than females, or vice versa, depending on operational sex-ratios). In sex changing species the
345 partitioning of reproductive success depends on the age at sex change. Some individuals change sex
346 earlier and might present higher overall lifetime reproductive success than those changing sex at older
347 stages, although shifts in age-at-sex-change will likely be linked with trade-offs with growth, maturity
348 and longevity (Allsop and West, 2003). The next step towards offering generality to this indication will
349 require a detailed understanding of how age-at-sex and other life-history traits shape the reproductive
350 success and N_e . Furthermore, the remarkably diverse magnitude range of N_e estimates obtained through
351 different methods, using the same data sets, suggests that, while these can be useful in a within-method
352 comparative approach, much remains to be done to match method estimation with the appropriate time
353 scales (Waples, 2005). We expect that the analysis of genetic data in larger sets of sex-changing and
354 gonochoristic species pairs will be required to shed more light on the significance of sex change in
355 affecting N_e and the evolutionary trajectories of populations and species.

356

357

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362

363 **Data Archiving**

364 Microsatellite and mitochondrial DNA data:

365 DRYAD entry XXXX → upon acceptance of manuscript.

366

367 **Conflict of interests**

368 The authors declare no conflict of interest.

369

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588

589 **FIGURE TITLES AND LEGENDS**

590

591

592 **Figure 1** – MAP: Sampling locations off the KwaZulu Natal Coast of South Africa.

593

594

595 **Figure 2** – EFFECTIVE POPULATION SIZE ESTIMATES: Effective population size of
596 *Cheimerius nufar* (santer, in red) and *Chrysoblephus puniceus* (slinger, in green) estimated with
597 different allelic frequencies (or P_{crit}). The three sampled populations were pooled.

598

599

600 **Figure 3** – HISTORICAL FEMALE EFFECTIVE POPULATION SIZE: Historical female effective
601 population size (N_f) of slinger, *Chrysoblephus puniceus* (continuous line) and santer, *Cheimerius nufar*
602 (dotted line) based on θ for different mutation rates (μ) and three generation lengths ($T = 3, 5$ and 8
603 years). Red values correspond to N_f estimations for $\mu = 11\%$ and $T = 5$.

604

605

606 **Figure 4:** BAYESIAN SKYLINE PLOTS: Bayesian Skyline plots for santer, *Cheimerius nufar*
607 and slinger, *Chrysoblephus puniceus*. The continuous and dashed lines represent the median and
608 mean values, respectively.

609

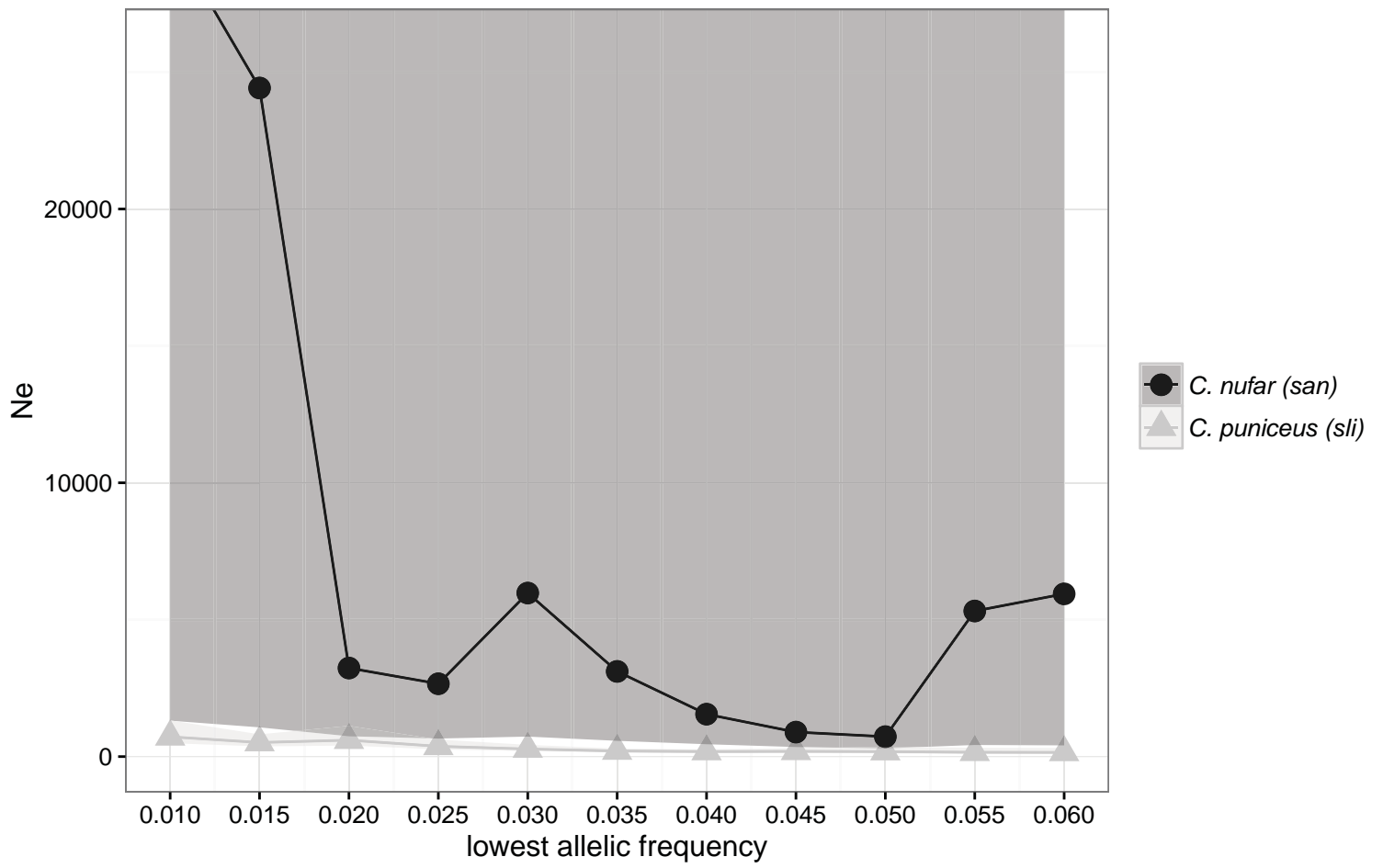
Table 1: Variability of ten polymorphic microsatellite loci in santer, *Cheimerius nufar*, from South Africa (n = 131) grouped into two multiplex reactions (I and II). Ta, annealing temperature (°C); Na, number of alleles observed; Ho, observed heterozygosity; He, expected heterozygosity. For slinger, *Chrysoblephus puniceus*, see Chopelet et al. (2009a).

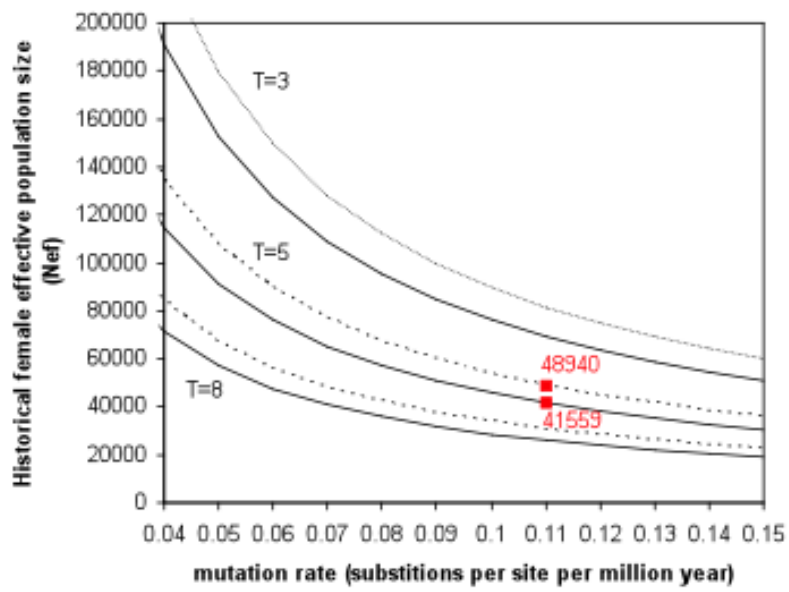
Locus name	GenBank Accession Number	Dye label	Primer sequences (5'-3')	Repeat Motif	Ta	Size Range (bp)	Na	He	Ho
SA1 (I)		6FAM	F:CAGCGATGCACAGTAAAGTACC R:AGCATACAGAGGCCTTCAGC	(TG) ₃₃	58	253-293	19	0.93	0.94
SA2 (II)		NED	F:GAGCCAGACTCCAGACATCC R:CCGGACAGGAGTATTGAAGC	(GTCT) ₁₁	58	190-234	8	0.75	0.73
SA3 (I)		PET	F:CCAGAGTCTGTGCTGTGAGTGC R:TCCTTGTGGTCCACTTTACG	(CA) ₁₇	58	386-420	13	0.89	0.77
SA6 (I)		6FAM	F:AGCTGCTGCTCATCTCACG R:GCAGTGTTAACATCTTCGAATGC	(TG) ₁₂	58	187-209	9	0.75	0.79
SA10 (I)		6FAM	F:GAAGCCAAACGAGGACAGC R:GTGAGGAGCATGCTAATACCG	(GT) ₁₅ GA(GT) ₂₆ GAGTGAGTGA(GT) ₁₈	58	428-532	42	0.96	0.92
SA25 (I)		VIC	F:GGAGGAAATGAACCGATGG R:GCAGCTGGTCAATAGTGTGG	(TG) ₇ CA(TG) ₆	58	152-220	19	0.61	0.70
SL25 (II)	FJ526983	NED	F:GGTACTGTTTGGCCCTTGC R:GCCTGGTAATATGCCTGAGC	(GA) ₁₁ TG(GA) ₃ TA(GA) ₄ GTCA (GA) ₂ AA(GA) ₉ CA(GA) ₂ AA(GA) ₁₂	58	207-271	7	0.61	0.66
SL26 (I)	FJ526984	PET	F:TGAAGGTGCTGATGACTTTCC R:CAGTCCTGCCTCTGACTGG	(TC) ₂ TT(TC) ₉ GC(TC) ₄ TGCCTT (TC) ₅ GC(TC) ₇	58	239-247	5	0.38	0.32
SL27 (I)	FJ526985	VIC	F:CAGCCTCAGCTCATTTCTCC R:CCTGCCCTCCTGTAGATGC	(TG) ₅₅	58	183-201	6	0.35	0.32
SL34 (I)	FJ526988	VIC	F:GCGTGCACACTCTTACAGTACC R:TCGGATGTGCATCTCATAGG	(CA) ₁₇	58	321-361	19	0.84	0.85

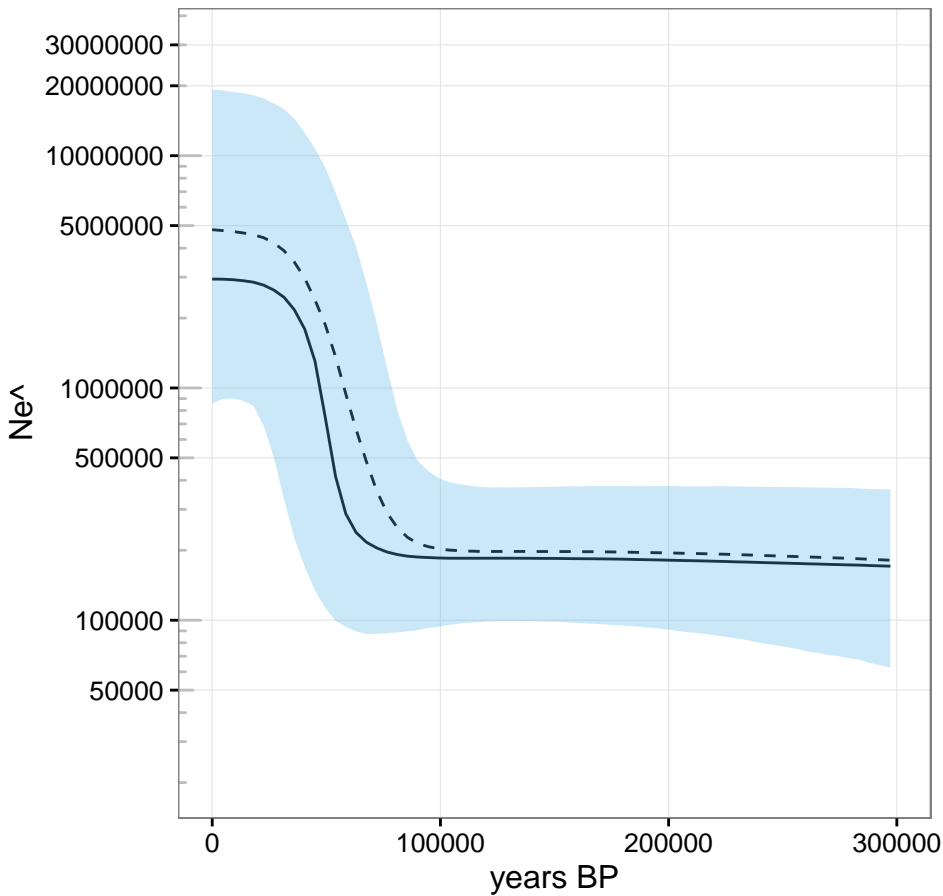
Table 2: Genetic diversity of slinger, *Chrysoblephus puniceus* and santer, *Cheimerus nufar* estimated with microsatellites and mitochondrial DNA. N, number of samples analysed; N_A , number of alleles; A_R , allelic richness; H_e , expected heterozygosity; H_o , observed heterozygosity; F_{is} , inbreeding coefficient; N_{hap} , number of haplotypes; H_d , haplotype diversity; π , nucleotide diversity. The three N_e estimates refer to, in order: i) estimates based on LDNe using all loci, ii) estimates based on LDNe excluding loci departing from HW equilibrium, iii) estimates based on DIYABC.

	location	microsatellites								mitochondrial				
		N	N_A	A_R	H_e	H_o	F_{is}	N_e^{\wedge}	$N_e^{\wedge(HWE)LD}$	$N_e^{\wedge(HWE)ABC}$	N	N_{hap}	H_d	p
<i>C. puniceus</i> <i>[slinger]</i>	PE	43	18.5	18	0.83	0.78	0.05	835(287- ∞)	707(259- ∞)		24	23	0.996	0.03
	PR	41	18.7	18.2	0.83	0.81	0.02	128(93-187)	535(228- ∞)		25	24	0.997	0.03
	RB	38	18	18	0.83	0.79	0.04	166(114-292)	462(198- ∞)		26	26	1	0.03
	ALL	125	18.4	18.1	0.83	0.79	0.04	371(289-511)	601 (404-1124)	560,000 (52,000-980,000)	75	65	0.996	0.03
<i>C. nufar</i> <i>[santer]</i>	PE	43	11.6	11.8	0.72	0.74	0.06	$\infty(-261-\infty)$	$\infty (327-\infty)$		25	22	0.99	0.04
	PR	45	12.4	12.4	0.74	0.76	-0.04	$\infty(-340-\infty)$	$\infty (733-\infty)$		26	20	0.969	0.04
	RB	42	12.1	12.1	0.72	0.70	0.02	$\infty(405-\infty)$	$\infty (353-\infty)$		26	24	0.994	0.04
	ALL	131	12	12.1	0.73	0.73	0.01	2236(654-∞)	3233 (743-∞)	370,000 (23,000-960,000)	77	55	0.984	0.04







C. nufar***C. puniceus***