Local parasite pressures and host genotype modulate epigenetic diversity in a mixed-mating fish

Waldir M. Berbel-Filho1 | Carlos Garcia de Leaniz1 | Paloma Morán2 | Joanne Cable3 | Sergio M. Q. Lima4 | Sofia Consuegra1

1Department of Biosciences, Swansea University, Swansea, UK
2Facultad de Biología, University of Vigo, Campus Universitario Lagoas-Marcosende, Vigo, Spain
3School of Biosciences, Cardiff University, Cardiff, UK
4Laboratório de Ictiologia Sistemática e Evolutiva, Departamento de Botânica e Zoologia, Universidade Federal do Rio Grande do Norte, Natal, Brazil

Correspondence
Sofia Consuegra, Department of Biosciences, Swansea University, Swansea SA2 8PP, UK.
Email: s.consuegra@swansea.ac.uk

Funding information
Conselho Nacional de Desenvolvimento Científico e Tecnológico, Grant/Award Number: 233161/2014-7; National Geographic/Waitt Foundation Grant, Grant/Award Number: W461-16

Abstract
Parasite-mediated selection is one of the main drivers of genetic variation in natural populations. The persistence of long-term self-fertilization, however, challenges the notion that low genetic variation and inbreeding compromise the host's ability to respond to pathogens. DNA methylation represents a potential mechanism for generating additional adaptive variation under low genetic diversity. We compared genetic diversity (microsatellites and AFLPs), variation in DNA methylation (MS-AFLPs), and parasite loads in three populations of Kryptolebias hermaphroditus, a predominantly self-fertilizing fish, to analyze the potential adaptive value of DNA methylation in relation to genetic diversity and parasite loads. We found strong genetic population structuring, as well as differences in parasite loads and methylation levels among sampling sites and selfing lineages. Globally, the interaction between parasites and inbreeding with selfing lineages influenced DNA methylation, but parasites seemed more important in determining methylation levels at the local scale.

KEYWORDS
epigenetic variation, hermaphroditism, inbreeding, local adaptation, mangrove killifish, self-fertilization

1 | INTRODUCTION

Organisms with mixed-mating reproduction (alternating between self-fertilization and outcrossing) can benefit from the advantages of both biparental and uniparental reproduction: outcrossing generates genetic variability and adaptability potential, while selfing ensures reproduction without partners (Jarne and Chalesworth 1993), and reproductive assurance (Darwin, 1876) gives self-reproducing individuals an advantage when colonizing new environments (Baker, 1955). The downside of selfing, however, is that the progeny can have reduced fitness compared to their outcrossed counterparts, usually suffering from inbreeding depression (Charlesworth & Willis, 2009). Thus, occasional outcrossing should be beneficial when inbreeding can impair offspring fitness (Damgaard, Couvet, & Loeschcke, 1992; Maynard Smith 1978).

The Red Queen hypothesis (Bell, 1982; Van Valen, 1973) is often invoked to explain the occurrence of sexual reproduction in face of the advantages of asexual reproduction (Bilr & Bell, 1987; Lively, 1987; Lively & Moran, 2014). According to this hypothesis, the more genetically diverse offspring of sexually reproducing individuals provide a "moving target" to parasites, making it more difficult for them to adapt compared to the "more static" offspring of asexual/uniparental individuals (Maynard Smith 1978; Hamilton, 1980; Lively, Craddock, & Vrijenhoek, 1990). Yet, while sexual reproduction seems the general rule in animals (approximately 99%; Slowinski et al., 2016), asexual and self-fertilizing lineages sometimes persist in...
with low levels of heterozygosity display high levels of genome-wide DNA methylation variation (Liebl, Schrey, Richards, & Martin, 2013; Richards, Schrey, & Pigliucci, 2012; Schrey et al., 2012), suggesting that DNA methylation could contribute to the adaptation of organisms with limited genetic diversity to environmental change (Castonguay & Angers, 2012; Douhovnikoff & Dodd, 2015; Liebl et al., 2013; Schrey et al., 2012; Verhoeven & Preite, 2014).

Increasing evidence suggests that epigenetic mechanisms, including genome-wide DNA methylation, are involved in host-pathogen interactions (Gómez-Díaz, Jordá, Peinado, & Rivero, 2012; Hu, Pérez-Jvostov, Blondel, & Barrett, 2018), but the mechanisms are better known in plants than in animals (Annacondia, Mageroy, & Martinez, 2018; Gómez-Díaz et al., 2012; Hewezi, Pantalone, Bennett, Neal Stewart, & Jr., Burch-Smith TM, 2018). Pathogenic infection in plants can result in hypomethylation of resistance-related genes but in hypermethylation at genome-wide level (Peng & Zhang, 2009). Mixed-mating organisms represent ideal models to test the associations between genetic and epigenetic variation with pathogen pressures because selfed and outcrossed offspring can naturally coexist, usually displaying very different levels of genetic diversity. Negative associations between genetic diversity and parasite loads have been previously observed in mixed-mating animals (Ellison, Cable, & Consuegra, 2011; Lively & Morran, 2014), with inbred individuals usually harboring more parasites. The relationship between epigenetic variation, parasites, and mixed-mating, however, has not been explored.

Here, we compared genetic diversity, variation in DNA methylation, and parasite loads in three natural populations of the mixed-mating mangrove killifish *Kryptolebias hermaphroditus* distributed along the Brazilian coast (Tatarenkov et al., 2017). The genus *Kryptolebias* contains the only known mixed-mating vertebrate species (*K. marmoratus* and *K. hermaphroditus*), characterized by variable rates of selfing and outcrossing (Tatarenkov et al., 2017). Populations of both species consist mainly of self-fertilizing hermaphrodites and varying levels of males at low frequencies (Berbel-Filho, Espirito-Santo, & Barbosa, 2016), suggesting that self-fertilization is the most common mode of reproduction (Avise & Tatarenkov, 2015).

We analyzed microsatellites (previously shown to correlate with parasite loads in the closely related *K. marmoratus*, see Ellison et al., 2011) and genome-wide methylation based on identification of anonymous CpG by methylation-sensitive AFLP (MS-AFLPs, previously used in nonmodel organisms) to identify epigenetic variation associated with parasite loads (Wenzel & Piertney, 2014). Based on the Red Queen hypothesis and previous results in *K. marmoratus*, we expected lower genetic diversity and higher parasite loads in inbred compared to outbred individuals. Given the relationship between genetic background and DNA methylation levels, we expected different patterns of variation in DNA methylation across selfing lines and predicted higher levels of DNA methylation in relation to inbreeding and parasite loads, if methylation played an adaptive role, potentially related to pathogen infection, in *K. hermaphroditus*.
2 | METHODS

2.1 | Study system, field sampling, and parasite screening

A total of 128 specimens of *K. hermaphroditus* were collected using hand-nets from three sampling sites on isolated mangroves on the northeastern coast of Brazil between January and September 2015: Ceará-Mirim River—Site 1; Curiimatáu River—Site 2; Ipojuca River—Site 3 (Figure 1). *K. hermaphroditus* is distributed along the Brazilian coast (Tatarakoven et al., 2017) and is typically found in shallow pools of high salinity levels (>30 ppt), clear waters, and muddy substrates, where there are few other sympatric fish (Berbel-Filho et al., 2016; Lira, Paiva, Ramos, & Lima, 2015). All specimens displayed the common hermaphroditic phenotype (dark color with well-defined ocellus on the caudal fin; Costa, 2011). Fish were euthanized using an overdose of tricaine methane sulfonate (MS-222) following UK Home Office Schedule 1 (Hollands, 1986), standard length was measured using a digital calliper (mm), and the whole fish were preserved in 95% ethanol at –20°C for parasite screening and DNA extraction.

In the laboratory, fish were dissected and screened for both external and internal parasite infections using a dissecting microscope following the methods of Ellison et al. (2011). Macroscopic parasite analyses focused on the three most common types of parasites identified. To assess the reliability of parasite screening, a subsample of five fish was examined by a different observer and the agreement was 100%. We defined parasite loads using a scaled measure of parasite abundance, where for each parasite morphotype (i), the number of parasites per individual (Ni) was divided by the maximum number found across all individuals (Nimax). The final value of the scaled parasite load represents the sum of scaled parasite loads across all parasite types. Given their uneven abundance (Table 1), this approach minimizes bias when parasite loads are heavily influenced by a very abundant parasite type (in our case, bacterial cysts) (Bolnick & Stutz, 2017).

2.2 | Genetic analysis

Genomic DNA from all 128 fish was extracted from gill tissue using a Nexttec extraction kit for blood and tissue samples (Nexttec, Leverkusen, Germany). Gills are an important physical and immunological barrier to pathogens in fish (Press & Evensen, 1999) and the organ where most parasites were found (Table 1). Twenty-seven microsatellite loci (Mackiewicz et al., 2006; Tatarakoven et al., 2017) were genotyped as in Ellison et al. (2011) and screened using GeneMapper v. 4.0 (Applied Biosystems). Loci were tested for linkage disequilibrium and Hardy–Weinberg equilibrium using GENEPOP v. 4.5.1 (Rousset, 2008). Mean number of alleles per locus (Na), observed heterozygosity (Ho), and expected heterozygosity (He) were estimated using GenALEX v. 6.5 (Peakall & Smouse, 2012). The inbreeding coefficient (Fis) was calculated in GENEPOP. Global heterozygosity for individual fish was estimated using the homozygosity by locus index (HL) implemented in the Excel macro Cernicalin v 1.3 (Aparicio, Ortego, & Cordero, 2006).

### TABLE 1 Genetic diversity (at 27 microsatellite loci), mean parasite number (standard deviation in brackets), and parasite prevalence in *Kryptolebias hermaphroditus* at sampling sites in northeastern Brazil

|                      | Site 1 | Site 2 | Site 3 | All sites |
|----------------------|--------|--------|--------|-----------|
| **Genetic diversity**|        |        |        |           |
| N                    | 68     | 42     | 18     | 128       |
| Na                   | 3.03   | 3.44   | 3.14   | 3.21      |
| Ho                   | 0.28   | 0.26   | 0.33   | 0.295     |
| He                   | 0.025  | 0.015  | 0.043  | 0.028     |
| FIS                  | 0.91   | 0.94   | 0.87   | 0.93      |
| HL                   | 0.95   | 0.97   | 0.93   | 0.95      |
| S                    | 0.92   | 0.93   | 0.87   | 0.90      |
| **Parasite loads**   |        |        |        |           |
| Bacterial gill cysts | 3.16 (3.16) | 2.66 (3.10) | 1.27 (0.80) | 2.73 (2.99) |
| Protozoan gill cysts | 0      | 1.52 (1.60) | 0.33 (1.37) | 0.54 (1.26) |
| Nematodes            | 0.16 (0.53) | 0.02 (0.15) | 0      | 0.09 (0.40) |
| **Total parasite load** | 3.33 (3.27) | 4.21 (3.17) | 1.61 (1.73) | 3.38 (3.17) |
| **Parasite prevalence (% of fish with infection)** |        |        |        |           |
| Bacterial gill cysts | 91.17 | 71.42  | 83.33  | 83.59     |
| Protozoan gill cysts | 0     | 57.14  | 5.55   | 19.53     |
| Nematodes            | 10.29  | 2.38   | 0      | 6.25      |

Abbreviations: N, sampling size; Na, mean number of alleles of alleles; Ho, observed heterozygosity; He, expected heterozygosity; FIS, inbreeding coefficient; HL, homozygosity by locus; S, selfing rates.

We also used the Bayesian clustering algorithm INSTRUCT (Gao, Williamson, & Bustamante, 2007) to estimate the optimal number of selfing lineages (k) with four simultaneous chains of 2,000,000 MCMC runs, 10 as thinning, and 100,000 of burn-in period, resulting in 100,000 interactions for each chain. The potential number of k tested ranged from 2 to 12. We used the individual q-values (the likelihood of membership to a particular genetic cluster or selfing lineage) from INSTRUCT to classify individuals as either selfed or outcrossed (Vähä & Primmer, 2006). A threshold of q-value ≥ 0.9 was used to classify selfed individuals, while <0.9 represented hybrids between two different selfing lineages, suggesting an outcrossing event (Ellison et al., 2011; Vähä & Primmer, 2006). Pairwise FST values among sampling sites and selfing lineages were estimated with Arlequin v. 3.5.2.2 (Excoffier & Lischer, 2010) using 10,000 permutations. We used hierarchical analysis of molecular variance (AMOVA) to investigate population structuring among sampling sites and selfing lineages (according to individual q-values) using 10,000 randomizations. Differences between selfed and outcrossed groups in the total number of parasites and homogygosity by locus (microsatellites) were analyzed using median Mann–Whitney rank tests in R v. 3.3.
2.3 Epigenetic analysis

We used methylation-sensitive amplified fragment length polymorphisms (MS-AFLPs) to assess genome-wide DNA methylation patterns (Schrey et al., 2013). DNA extracted from gill filament tissue of 115 fish (33 classified as outcrossed and 82 as selfed according to the INSTRUCT q-values; 62, 36, and 17 from sampling sites 1, 2, and 3, respectively) was used for the MS-AFLP analysis following Rodríguez López et al. (2012). A DNA aliquot of 100 ng per individual was split for digestion with two enzyme combinations: EcoRI/HpaII and EcoRI/Mspl. The digested DNA was ligated to adaptors, and a selective PCR was conducted using the primers ECORI- ACT: GACTGCGTACCAATTCACT and HPA- TAG: GATGAGTCTAGAACGGTAG following Ellison et al. (2015). The HpaII primer was end-labeled with 6-FAM. Fragments were run on an ABI PRISM 3100 (Applied Biosystems), and the resultant profiles were analyzed using GENEMAPPER v. 4.0 (Applied Biosystems). To ensure reproducibility, the following settings were applied: Analysis range was 100–500 bp; minimum peak height was 100 relative fluorescence units; pass range for sizing quality was 0.75–1.0; and maximum peak width was 1.5 bp. To confirm MS-AFLP reproducibility, 24 individuals (~20% of the total; eight from each sampling site) were reanalyzed and compared using the same protocols.

The R package msap v. 1.1.9 (Pérez-Figueroa, 2013) was used to analyze MS-AFLP data. To increase reproducibility of the genotyping, we used an error threshold of 5% as suggested by Herrera and Bazaga (2010). According to the binary band patterns, each locus was classified as either methylation-susceptible loci (MSL; i.e., displaying a proportion of HPA+/MSP− and/or HPA−/MSP+) or nonmethylated loci (NML; if the same patterns did not exceed the error threshold) (Pérez-Figueroa, 2013). MSL were used to assess epigenetic variation, while NML were used as a measure of AFLP genetic variation. Average group methylation percentages for inbreeding status were calculated using the different binary band combinations: EcoRI/HpaII and EcoRI/Mspl. The digested DNA was ligated to adaptors, and a selective PCR was conducted using the primers ECORI- ACT: GACTGCGTACCAATTCACT and HPA- TAG: GATGAGTCTAGAACGGTAG following Ellison et al. (2015).

The R package msap v. 1.1.9 (Pérez-Figueroa, 2013) was used to analyze MS-AFLP data. To increase reproducibility of the genotyping, we used an error threshold of 5% as suggested by Herrera and Bazaga (2010). According to the binary band patterns, each locus was classified as either methylation-susceptible loci (MSL; i.e., displaying a proportion of HPA+/MSP− and/or HPA−/MSP+) or nonmethylated loci (NML; if the same patterns did not exceed the error threshold) (Pérez-Figueroa, 2013). MSL were used to assess epigenetic variation, while NML were used as a measure of AFLP genetic variation. Average group methylation percentages for inbreeding status were calculated using the different binary band combinations: EcoRI/HpaII and EcoRI/Mspl. The digested DNA was ligated to adaptors, and a selective PCR was conducted using the primers ECORI- ACT: GACTGCGTACCAATTCACT and HPA- TAG: GATGAGTCTAGAACGGTAG following Ellison et al. (2015).

Epigenetic (MSL) and genetic (NML) differentiation at AFLPs among sampling sites, selfing lineages, and between outcrossed and selfed groups was assessed by AMOVA with 10,000 randomizations. Epigenetic (MSL) and genetic (AFLP and microsatellites) differentiation among sampling sites, selfing lineages, and inbreeding status was visualized by principal coordinates analysis (PCoA). Mantel tests based on distance matrices (Mantel, 1967) were used to test for potential correlations between epigenetic and genetic data for MSL, NML, and microsatellites using GÉNALEX v. 6.5 with and 10,000 permutations. To identify disproportionately differentiated methylation states, we used a FST outlier approach implemented in BayeScan 2.1 (Foll & Gaggiotti, 2008; Perez-Figueroa et al., 2010), with 2 × 106 iterations (thinning interval 20 after 20 pilot runs of 104 iterations each) and a burn-in of 5 × 105. We tested for outliers based on the MSL data generated on the comparisons among sampling sites, selfing lineages, and between inbreeding status (inbred or outbred).

2.4 Statistical analyses

A Kruskal–Wallis test was used to examine the differences on scaled parasite load and bacterial cysts (the most prominent parasite) among selfing lineages. To test the relationship between genome-wide variation in DNA methylation and parasite loads, the proportion of methylated loci per individual was calculated as the proportion of loci scored as methylated over the total number of loci observed per individual (“0” for unmethylated and “1” for methylated, excluding the missing data cells per individual). The proportion (or percentage) of methylated loci has been previously used to analyze differences in epigenetic profiles among groups (Ardua, Zaito, Morán, Planes, & Garcia-Vazquez, 2017; Groot, Wagemaker, Ouborg, Verhoeven, & Vergeer, 2018; Veerger et al., 2012), and has shown both inter- and intraspecific variation (Alonso, Pérez, Bazaga, Medrano, & Herrera, 2016).

Model selection was conducted using the multimodel averaging approach implemented in the R package glmulti v 1.0.7 (Calcagno & de Mazancourt, 2010). We chose the minimal adequate models based on the lowest AICc values (Akaike information criterion corrected for small sample size), Akaike weight (W.), and evidence ratios (Burnham & Anderson, 2004). Models (within 2 AIC units) were also reported. Predictors were checked for collinearity using pair.panels function in R package psych (Revelle et al., 2019). Model residuals were checked and assumptions validated.

To disentangle potential confounding effects arising from the unequal distribution of selfing lineages among sampling sites (i.e., five lineages are exclusive to a particular sampling site, Table S1), we repeated the analyses (AMOVA, Mantel test, PCoA, and GLMs) for both genetic (microsatellites and AFLPs) and epigenetic (MSL) data using only individuals from Site 1 (68 individuals for microsatellites and 62 for MS-AFLPs), as this was the only site with more than two selfing lineages (Table S1).

3 RESULTS

3.1 Parasite screening

Bacterial cysts were present on the gills and consisted of white to yellow spherical cysts circumscribed by a capsule, which resulted in hypertrophied gill filaments. They were the most common type of pathogen appearing in 83.6% of the individuals screened, with a
prevalence ranging from 1 to 19 (mean = 2.73, SD = 2.99), and were more prevalent in Site 1 (mean = 3.16, SD = 3.16), followed by Site 2 (mean = 2.66, SD = 3.10) and Site 3 (mean = 1.27, SD = 0.80). The second most common macroscopic parasites were protozoan cysts, which consisted of small dark oval cysts over the gill arch and filaments. In total, 19.53% of the total number of individuals were infected with these cysts, ranging from 1 to 6 (mean = 0.54, SD = 1.26). Protozoan cysts were absent in Site 1, but present in Site 2 (mean 1.52, SD = 1.6) and Site 3 (mean = 0.33, SD = 1.37). Finally, adult nematodes were found in the gut of only eight individuals (6.25%), ranging from 1 to 3 (mean = 0.09, SD = 0.40). Nematodes were only detected in Sites 1 (mean = 0.3, SD = 1.37) and 2 (mean = 0.02, SD = 0.15) (Figure S1; Tables 1 and S1). Only seven individuals (5.4%) were uninfected with macroparasites. Significant differences were found on scaled parasite loads (chi-square = 32.14, p ≤ 0.001, df = 5) and bacterial cysts (chi-square = 12.98, p = 0.01, df = 5) among selfing lineages.

3.2 Genetic diversity and population structuring based on microsatellites

No linkage disequilibrium was detected between any pair of microsatellite loci. As expected from the high levels of self-fertilization of the species, no loci were found to be in Hardy–Weinberg equilibrium, and all 27 microsatellite loci showed an excess of homozygotes. The global homozygosity index (HL) was very high (mean = 0.95), as well the estimated selfing rates (Table 1). At the individual level, 93 individuals (72.6%) were homozygous across all 27 microsatellite loci. However, 17 individuals (13.28%) displayed intermediate to high levels of heterozygosity (ranging from 0.13 to 0.69).

The clustering Bayesian algorithm INSTRUCT indicated that six was the most likely number of selfing lineages (k). Selfing lineage 6 was shared between two different mangroves (Site 1 with seven individuals and Site 2 with one individual), separated by approximately 100 km. The other five lineages were solely represented in one of the mangroves (lineage 1 with 14 individuals, lineage 2 with 25 individuals, and lineage 4 with 22 individuals in Site 1; lineage 3 with 41 individuals in Site 2; and lineage 5 with 18 individuals in Site 3) (Figures 1 and 2; Table S1). High FST values were found both among sampling sites (mean = 0.28, SD = 0.02) and selfing lineages (mean = 0.32, SD = 0.05). All pairwise comparisons were highly significant (Table S2).

Based on the results from the INSTRUCT analysis, the fish were classified as selfed or outcrossed on the basis of their q-values (Vaha & Primmer, 2006), following an approach previously used in the also mixed-mating K. marmoratus (Ellison et al., 2011). On this basis, 92 fish (71%; 46 from Site 1, 30 from Site 2, and 16 from Site 3) were classified as selfed (with q-values ≥ 0.9) and 36 (29%; 22 from Site 1, 12 from Site 2, and two in Site 3) as outcrossed (with q-values < 0.9) (Figure 2; Table S1). The classification of individuals as selfed or outcrossed is based on the lineage composition; hence, homozygote individuals can be classified as originated from outcrossing if they display alleles from different lineages, even if they appear in homozygosity after several generations of selfing. Overall, outcrossed individuals had significantly lower homozygosity by locus values (at microsatellites) and total parasite loads than selfed individuals (Table 2).

Overall, AMOVA using microsatellites indicated strong and significant differentiation among sampling sites (FST = 0.28, p = 0.001) and selfing lineages (FST = 0.32, p = 0.001) (Table 3). Although significant, very low genetic differentiation was found between selfed and outcrossed individuals (FST = 0.01, p = 0.002) (Table 3; Figure S2). These patterns were also seen on PCoA, with individuals generally clustering by selfing lineages in the microsatellites data (25.84% of overall variation), with individuals from lineage 4 being the most differentiated from the other lineages on Site 1. In this site, substantial overlap was found among selfing lineages and between selfed and outcrossed, despite its significant differences (FST = 0.03, p = 0.001) (Table S4; Figure S3).

3.3 Genetic and epigenetic variability and population structuring based on MS-AFLPs

The epigenetic analysis identified 381 MS-AFLP loci, of which 267 (70.07%) were methylation-susceptible loci (MSL) and 106
BERBEL‐FILHO et al.

(27.82%) nonmethylated loci (NML). Of the MSL loci, 236 (88.3%) were polymorphic and therefore were used for the variability analysis. Reproducibility comparisons between original and replicated genotypes for 24 individuals revealed 238 loci with an average of 0.5% error rate (differences across individuals divided by the number of loci times number of replicates, as in Bonin et al., 2004), which is within the normal reproducibility range for AFLPs genotyping (Bonin et al., 2004). AMOVA for reproducibility also revealed no significant differences between methylation and AFLP variation patterns between original and replicated set of individuals (Table S3). Average methylation ranged from 47.51% on lineage 2 to 38.17% on lineage 5, and was 44.82% for inbred and 45.77% for outbred individuals.

AMOVA revealed very low but significant differentiation among sampling sites, for both genetic (AFLPs: \( \phi_{ST} = 0.02 \), \( p = 0.001 \)) and epigenetic (\( \phi_{ST} = 0.02 \), \( p < 0.001 \)) loci. Significant differentiation among selfing lineages was also found on genetic (AFLPs: \( \phi_{ST} = 0.02 \), \( p = 0.004 \)) and epigenetic (\( \phi_{ST} = 0.02 \), \( p = 0.001 \)) loci. Overall, higher genetic and epigenetic variance was found within than between groups (Table 3). As with microsatellites, no clear genetic (at AFLPs) or epigenetic differentiation was found between selfed and outcrossed individuals (Figure S2).

There was, however, a significant positive association between epigenetic (MSL) and genetic diversity, both using AFLPs (Mantel test, \( r = 0.11 \); \( p = 0.002 \)) and microsatellites (\( r = 0.09 \); \( p = 0.001 \)). No MSL epiloci were identified as an \( F_{ST} \) outlier in any of the comparisons.

---

**TABLE 2** Comparison of homozygosity by locus (HL) (at 27 microsatellite loci), mean parasites loads (standard error in brackets), and parasite prevalence between *Kryptolebias hermaphroditus* classed as either selfed or outcrossed based on q-values from selfing lineage structure estimated using INSTRUCT

|                | Selfed | Outcrossed | \( z \)   | \( p \) value |
|----------------|--------|------------|-----------|---------------|
| Genetic diversity |        |            |           |               |
| N              | 92     | 36         |           |               |
| HL             | 0.98   | 0.88       | -4.76     | <0.001        |
| Parasite loads |        |            |           |               |
| Bacterial gill cysts | 3.25 (2.99) | 1.69 (2.59) |           |               |
| Protozoan gill cysts | 0.57 (1.26) | 0.47 (1.28) |           |               |
| Nematodes      | 0.1 (0.4) | 0.05 (0.42) |           |               |
| Total parasite load | 3.82 (3.47) | 2.25 (1.94) | -2.84      | 0.004         |
| Parasite prevalence (% of fish with infection) |        |            |           |               |
| Bacterial gill cysts | 89.13  | 69.44      |           |               |
| Protozoan gill cysts | 18.47  | 22.22      |           |               |
| Nematodes      | 7.6    | 2.77       |           |               |

Note: \( p \) and \( z \)-values extracted from a two median Mann–Whitney test.

---

**TABLE 3** Hierarchical analysis of molecular variance (AMOVA) for microsatellites and MS-AFLPs data among (a) sampling sites, (b) selfing lineages, and (c) selfed and outcrossed individuals in *Kryptolebias hermaphroditus*

|                | Microsatellites | MS-AFLPs |
|----------------|----------------|----------|
|                | NML            | MSL      |
| df             | Mod. var. (%)  | \( p \) value |
| Sampling sites |                |          |
| Among sites    | 2 28.46        | 0.28     |
| Within sites   | 227 71.54      | 0.15     |
| Within lineages| 5 32.40        | 0.32     |
| Between selfed and outcrossed | 1 2.84      | 0.01 |
| df             | Mod. var. (%)  | \( p \) value |
| Among sites    | 2 2.20         | 0.02     |
| Within sites   | 112 97.80      | 0.02     |
| Within lineages| 5 2.00         | 0.02     |
| Between selfed and outcrossed | 1 0.82      | 0.02 |
| df             | Mod. var. (%)  | \( p \) value |
| Among sites    | 2 2.96         | 0.02     |
| Within sites   | 112 97.85      | 0.02     |
| Within lineages| 5 2.15         | 0.02     |
| Between selfed and outcrossed | 1 0.82      | 0.02 |
| df             | Mod. var. (%)  | \( p \) value |
| Among sites    | 2 2.15         | 0.02     |
| Within sites   | 109 98.00      | 0.02     |
| Within lineages| 5 2.00         | 0.02     |
| Between selfed and outcrossed | 1 0.82      | 0.02 |

Note: \( p \) value derived from 10,000 permutations. Abbreviations: df, degrees of freedom; Mod. var. (%), molecular variance percentages from variance components \( \sigma^2 \); \( \phi_{ST} \), phi statistics for population differentiation; SSD, sum of squared deviations.
No significant differences between selfing lineages were found among lineages for individuals from Site 1 for AFLPs genetic data (selfing lineages: $\phi_{ST} = 0.008, p = 0.12$) or MSL epigenetic data (selfing lineages: $\phi_{ST} = 0.006, p = 0.20$) (Table S4). In the PCoA, substantial overlap was found among selfing lineages and between selfed and outcrossed individuals (Figure S3). Mantel tests between genetic and epigenetic data indicated a significant positive association between AFLPs and MSL data ($r = 0.21; p < 0.001$), but not between microsatellites and MSL ($r = -0.005; p = 0.45$).

### 3.4 Parasite loads, genetic variation, and epigenetic variation

According to a multimodel testing approach, the most plausible model for the proportion of methylated DNA included selfing lineage, scaled parasite load, inbreeding status, and the interactions between selfing lineage and scaled parasite load and inbreeding. The proportion of methylated loci significantly varied among selfing lineages (estimate = 0.51, SE = 0.13, $p < 0.001$) and was affected...
by parasite loads and inbreeding status through its interactions with selfing lineage (parasite loads and selfing lineage: estimate = −0.55, SE = 0.46, p = 0.005; inbreeding and selfing lineage interaction: estimate = −1.64, SE = 0.14, p = 0.04) (Figure 3b–c; Tables 4 and S7). The second most likely model (ΔAICc = 1.00) included only selfing lineage (estimate = −0.43, SE = 0.08, p < 0.001) and the interactions between inbreeding and selfing lineage (estimate = −1.10, SE = 0.12, p = 0.04) as significant predictors. However, this model explained substantially less of the overall variation compared to the first model (weight: 0.17 vs. 0.28) and was 1.39 times less likely than the first one (Tables S5–S6).

Overall, the results of the single-taxon models (using number of bacterial cysts) were very similar to those for scaled parasite loads. The best model to explain the proportion of methylated loci included selfing lineage, and the interactions between selfing lineage and bacterial cysts, and selfing lineage and inbreeding (Table S7).

When using only individuals from Site 1 (to remove any potential confounding effect between sampling site and selfing lineages) for the proportion of methylated loci, the model with the lowest AIC indicated that selfing lineage, inbreeding, and the interactions between inbreeding and selfing lineage and inbreeding and scaled parasite loads were all significant predictors (Table S8). However, the second best-fitting model (ΔAICc = 0.02) explained the same amount of variation (weight = 0.39) and the evidence ratio (0.66) suggested that it was more likely (evidence ratio of 1.50) than the first model. This second model indicated that overall, the proportion of methylated DNA significantly increased with scaled parasite loads (estimate = 0.43, SE = 0.11, p = 0.03) and that DNA methylation levels were also affected by the interaction between scaled parasite loads and inbreeding (estimate = −1.29, SE = 0.38, p < 0.001), with inbred individuals having increased methylation levels with increased parasite loads, while outbred individuals had decreased methylation levels with increased parasite loads (Figure 3d; Table 4).

4 | DISCUSSION

DNA methylation could play an important adaptive role in organisms with low heterozygosity, including self-fertilizing species, potentially increasing their plasticity capacity to cope with environmental change (Douhoukonkoff & Dodd, 2015; Verhoeven & Pretie, 2014). However, our results did not indicate significant differences in genome-wide DNA methylation variation between selfed and outcrossed individuals, and our models only identified inbreeding status (defined as originating from selfing or outcrossing) significantly related to DNA methylation via its interaction with selfing lineage (all sampling sites) and parasites (at the local scale in Site 1). Higher variation in DNA methylation has been reported for clonal and inbred individuals (Liebl et al., 2013; Massicotte & Angers, 2012; Nakamura & Hosaka, 2010; Richards et al., 2012; Veerger et al., 2012) and has been interpreted as an adaptive mechanism to compensate for low genetic variation (Schrey et al., 2012), or as a potential consequence of inbreeding (as in Verger, Wagemaker, & Ouborg, 2012) responsible, at least in part, for inbreeding depression (Nakamura & Hosaka, 2010; Verger et al., 2012). Yet, our results suggest that, at least in this species, either inbreeding does not affect genome-wide DNA methylation variation or it does in a gene-specific manner (Venney, Johansson, & Heath, 2016), although further research would be needed to address this question.

We found that the different selfing lineages of Kryptolebias hermaphroditus distributed in three sampling sites of northeastern Brazil differed significantly in parasite loads, genetic composition, and DNA methylation patterns, which might indicate specific interactions between host genotypes, epigenotypes, and parasites (Dybdahl & Lively, 1998; Ebert, 2008). Previous studies on mangrove killifishes had identified extensive genetic structuring between both (Tatarenkov et al., 2015, 2017) and within mangrove systems even at close geographical proximity (Ellison et al., 2012; Tatarenkov, Earley, Taylor, & Avise, 2012; Tatarenkov et al., 2007), as a consequence of the self-fertilizing nature of these fish. We found strong evidence of genetic structuring between sampling sites and selfing lineages using microsatellites, but lower differentiation for AFLP genetic markers (likely due to the different mutation rate of the markers) and epigenetic markers (MS-AFLPs).

Overall, more inbred individuals harbored higher parasite loads than their outcrossed counterparts, supporting the prediction that low heterozygosity due to self-fertilization may reduce fitness (considering parasite loads as a proxy for pathogen pressure), as for other mixed-mating species (Ellison et al., 2011; King, Jokela, & Lively, 2011; Lively & Morr, 2014). Extensive periods of self-fertilization can reduce offspring fitness due to the accumulation of deleterious alleles and inbreeding depression (Charlesworth et al. 1993). Species with mixed-mating seem to overcome these problems through occasional outcrossing (Ellison et al., 2011; Morr, 2014). Therefore, we propose that DNA methylation could play an additional role in these species, potentially increasing their plasticity capacity to cope with environmental change (Douhoukonkoff & Dodd, 2015; Verhoeven & Pretie, 2014).
Schmidt, Gelarden, Parrish, & Lively, (2011), which can generate genetic diversity to face natural enemies, such as parasites (Lively 2014). Here, the relationship between parasites and inbreeding status (selfed or outcrossed) suggests that outcrossing might confer a fitness advantage (in terms of parasite loads), even when it occurs at very low frequencies (Ellison et al., 2011). However, despite the adaptive potential of outcrossing, the main reproductive mode of K. hermaphroditus seems to be self-fertilization (Tatarenkov et al., 2017). This suggests that other evolutionary mechanisms may be balancing the harmful effects of parasite infections or that parasite selection is of low (Lively, 2014), as theory predicts that low selection levels imposed by natural enemies are consistent with the maintenance of asexual reproduction (Judson, 1997; Ladle, Johnstone, & Judson, 1993). For example, in the mixed-mating Potamopyrgus snails, the oldest asexual lineages are restricted to populations where parasites are rare (Neiman, Jokela, & Lively, 2005). Thus, the low number of parasites found in K. hermaphroditus (i.e., mean of 3.38 parasites per individual compared to 22.41 of K. marmoratus in Belize; Ellison et al., 2011) may explain the high prevalence of selfing in K. hermaphroditus.

The long-term persistence of self-fertilizing organisms suggests that nongenetic mechanisms may play a role in regulating gene expression to cope with environmental change (Douhovnikoff & Dodd, 2015; Hu et al., 2018; Liebl et al., 2013; Shrey et al., 2012). However, recent studies indicate that DNA methylation is likely to interact with genotypes in a genotype-by-environment manner to generate plastic responses (Herman & Sultan, 2016). For example, Dubin and colleagues (2015) found strong influence of genetic variants in DNA methylation levels in response to different temperature regimes in A. thaliana. In humans, either the genotype alone or genotype-by-environment interactions in the uterus explained the variation of over a thousand differentially methylated regions on the methylome of neonates (Teh et al., 2014). Using data from all sampling sites, we found that genome-wide DNA methylation was strongly influenced by selfing lineage, and only at a smaller scale by inbreeding through its interaction with selfing lineage (Bell et al., 2011; Bjorronsson et al., 2008; Dubin et al., 2015; Gertz et al., 2011). Strong epigenetic differences between selfing lines had been identified previously in K. marmoratus (Berbel-Filho et al., 2019; Ellison et al., 2015), indicating an important role of the genetic background in the epigenetic variation of mangrove killifishes. In addition, we also found a significant correlation between DNA methylation and genetic variation (at both AFLP and microsatellites data), suggesting that autonomous variation in DNA methylation may be limited (Dubin et al., 2015).

Several abiotic and biotic factors, including parasites (Hu et al., 2018; Norouzitallab et al., 2014) as well as stochastic epimutations (Leung et al., 2016), are known to influence DNA methylation variation. Our results showed that genome-wide DNA methylation levels for all sampling sites were significantly influenced by parasite loads through the interaction with selfing lineage, suggesting a potential genotype-by-environment interaction on parasite responses. Yet, as most of the selfing lineages were exclusive to specific sampling sites, we could completely discard confounding effects between both variables. In fact, selfing lineage did not affect genome-wide DNA methylation levels in Site 1, but only parasites and their interaction with inbreeding status. Increasing evidence has been showing that DNA methylation is involved in the modulation of host-pathogen interactions (Gómez-Díaz et al., 2012). The bacterial parasite Wolbachia, for example, alters host genome-wide DNA methylation patterns resulting in the feminization of infected leafhoppers (Zygindia pullulan) to increase its transmission (Negri et al., 2009). Experiments in plants with both hyper- and hypomethylated mutants indicate that genome-wide DNA demethylation enhances immune responses to both bacterial (Dowen et al., 2012; Yu et al., 2013) and fungal (Le et al., 2014) infections. Additionally, almost half of the resistance genes in the Arabidopsis genome are regulated by DNA methylation, which shows the importance of this pathway in the global regulation of resistance activation (López Sánchez et al., 2016). Although we found evidence of parasites affecting DNA methylation variation, the anonymous nature of our genetic and epigenetic markers is a limiting factor to infer the potential adaptive/functional role of the DNA methylation variation in response to parasites. Further analyses, ideally under controlled experimental conditions and using higher resolution sequencing methods (i.e., whole-genome bisulfite sequencing, RNAseq, CRISPR/Cas9), should help to clarify how DNA methylation may affect immune responses in mixed-mating Kryptolebias species.

The relationship between parasite loads and outcrossing seems to be common to several mixed-mating species (Ellison et al., 2011; King et al., 2011; Steets, Wolf, Auld, & Ashman, 2007) in addition to K. hermaphroditus, suggesting that the influence of parasites in the regulation of mixed-mating could be generalized. The extent of this relationship, however, may depend on the severity of the selection imposed by coevolving parasites (Lively & Morran, 2014). Our results indicate that genotype composition (and its interaction with inbreeding) may be important in DNA methylation responses to environmental variation in wild populations, and that, if DNA methylation responded in a genotypic-specific manner to parasites pressures, it could contribute to local adaptation (Foust et al., 2016; Smith, Mártiln, Nguyen, & Mendelson, 2016). The mangrove killifish, with its naturally inbred populations and marked methylation differences between populations and genotypes, represents an ideal model to analyze the relative roles of genetic and epigenetic diversity in modulating local adaptation.

ACKNOWLEDGMENTS

This work was supported by the Conselho Nacional de Desenvolvimento Científico e Tecnológico (CNPq) through a Science without Borders scholarship (233161/2014-7). This work was also funded by the National Geographic/Waitt Foundation Grant (W461-16). We thank Amy Ellison for the help with parasite screening. We are thankful for the LISE laboratory team and CTA facilities from Universidade Federal do Rio Grande do Norte for support during fieldwork.
AUTHORS’ CONTRIBUTIONS

SC, WMB-F, and CGL conceived the work; SMQL planned the field work and conducted the sampling together with WMB-F, CGL, and SC; WMB-F did the microsatellite and parasite screening, with contributions from JC; WMB-F and PM performed the MS-AFLP analyses. WMB-F analyzed the data with the contribution of SC, CGL, and PM. WMB-F and SC wrote the paper with contributions from all authors.

ETHICAL APPROVAL

All the experiments in this study have been conducted following Home Office regulations, approved by Swansea, Cardiff and UFRN (CEUA) Universities Animal Ethics Committees, and under sampling permit number 30532-1/2011 issued by ICMBio/SISBIO. The authors declare they have no conflict of interest.

DATA AVAILABILITY STATEMENT

Data are available from Dryad Digital Repository: 10.5061/dryad.0065k4k.

ORCID

Waldir M. Berbel-Filho https://orcid.org/0000-0001-6991-4685
Carlos Garcia de Leanz https://orcid.org/0000-0003-1650-2729
Paloma Morán https://orcid.org/0000-0002-3644-8507
Joanne Cable https://orcid.org/0000-0002-8510-7055
Sergio M. Q. Lima https://orcid.org/0000-0001-9365-4879
Sofia Consuegra https://orcid.org/0000-0003-4403-2509

REFERENCES

Alonso, C., Pérez, R., Bazaga, P., Medrano, M., & Herrera, C. M. (2016). MSAP markers and global cytosine methylation in plants: A literature survey and comparative analysis for a wild-growing species. Molecular Ecology Resources, 16, 80–90. https://doi.org/10.1111/1755-0998.12426
Annacondia, M. L., Mageroy, M. H., & Martinez, G. (2018). Stress response regulation by epigenetic mechanisms: Changing of the guards. Physiologia Plantarum, 162, 239–250. https://doi.org/10.1111/pla.12662
Aparicio, J. M., Ortego, J., & Cordero, P. J. (2006). What should we weigh to estimate heterozygosity, alleles or loci? Molecular Ecology, 15, 4659–4665. https://doi.org/10.1111/j.1365-294X.2006.03111.x
Ardura, A., Zaito, A., Morán, P., Planes, S., & García-Vazquez, E. (2017). Epigenetic signatures of invasive status in populations of marine invertebrates. Scientific Reports, 7, 42193.
Avise, J. C., & Tatarenkov, A. (2015). Population genetics and evolution of the mangrove rivulus Kryptolebias marmoratus, the world’s only self-fertilizing hermaphroditic vertebrate. Journal of Fish Biology, 87, 519–538. https://doi.org/10.1007/jfb.12741
Baker, H. G. (1955). Self-compatibility and establishment after “long-distance” dispersal. Evolution, 9, 347–349. https://doi.org/10.2307/2405656
Bell, G. (1982). The Masterpiece of Nature: The evolution and genetics of sexuality. Berkeley: University of California Press.
Bell, J. T., Pai, A. A., Pickrell, J. K., Gaffney, D. J., Pique-Regi, R., Degner, J. F., … Pritchard, J. K. (2011). DNA methylation patterns associate with genetic and gene expression variation in HapMap cell lines. Genome Biology, 12, R10. https://doi.org/10.1186/gb-2011-12-1-r10
Berbel-Filho, W. M., Espirito-Santo, H. M. V., & Lima, S. M. Q. (2016). First record of a male of Kryptolebias hermaphroditus Costa, 2011 (Cyprinodontiformes: Cyprinodontidae). Neotrop Ichthyol, 14, e160024. https://doi.org/10.1590/1982-0224-20160024
Berbel-Filho, W. M., Rodríguez-Barreto, D., Berry, N., García de Leanz, C., & Consuegra, S. (2019). Contrasting DNA methylation responses of inbred fish lines to different rearing environments. Epigenetics (just-accepted).
Bjorsson, H. T., Sigurðsson, M. I., Fallin, M., Irizarry, R. A., Asplund, T., Cui, H., … Feinberg, A. P. (2008). Intra-individual change over time in DNA methylation with familial clustering. JAMA, 299, 2877–2883. https://doi.org/10.1001/jama.299.24.2877
Blair, A., & Bell, G. (1987). Mammalian chiasma frequencies as a test for a two theories of recombination. Nature, 326, 803. https://doi.org/10.1038/326803a0
Bolnick, D. I., & Stutz, W. E. (2017). Frequency dependence limits divergent evolution by favouring rare immigrants over residents. Nature, 546, 285. https://doi.org/10.1038/nature22351
Bonduriansky, R., & Day, T. (2018). Extended Heredity: a new understanding of inheritance and evolution. Princeton: Princeton University Press.
Bonin, A., Bellemain, E., Bronken Eidesen, P., Pompanon, F., Brochmann, C., & Taberlet, P. (2004). How to track and assess genotyping errors in population genetics studies. Molecular Ecology, 13, 3261–3273. https://doi.org/10.1111/j.1365-294X.2004.02346.x
Bosdorf, O., Richards, C. L., & Pigliucci, M. (2008). Epigenetics for ecologists. Ecology Letters, 11, 106–115. https://doi.org/10.1111/j.1461-0248.2007.01130.x
Burnham, K. P., & Anderson, D. R. (2004). Multimodel inference - understanding AIC and BIC in model selection. Sociological Methods & Research, 33, 261–304. https://doi.org/10.1177/0049124104268644
Calcagno, V., & de Mazancourt, C. (2010). glmulti: An R package for easy automated model selection with (generalized) linear models. Journal of Statistical Software, 34, 1–29. https://doi.org/10.18637/jss.v034.i12
Castonguay, E., & Angers, B. (2012). The key role of epigenetics in the persistence of asexual lineages. Genetics Research International, 2012, 534289. https://doi.org/10.1155/2012/534289
Charlesworth, D., & Willis, J. H. (2009). The genetics of inbreeding depression. Nature Reviews Genetics, 10, 783–796. https://doi.org/10.1038/nrg2664
Costa, W. J. E. M. (2011). Identity of Rivulus ocellatus and a new name for a hermaphroditic species of Kryptolebias from south-eastern Brazil (Cyprinodontiformes: Rivulidae). Ichthyological Exploration of Freshwaters, 22, 185–192.
Damgaard, C., Couvet, D., & Loeschcke, V. (1992). Partial selfing as an optimal mating strategy. Heredity, 69, 289–295. https://doi.org/10.1038/hdy.1992.127
Darwin, C. (1876). Cross and self-fertilization of plants. London, UK: Murray.
Douhouvnikoff, V., & Dodd, R. S. (2015). Epigenetics: A potential mechanism for clonal plant success. Plant Ecology, 216, 227–233. https://doi.org/10.1007/s11258-014-0430-z
Down, R. H., Pelizzola, M., Schmitz, R. J., Lister, R., Down, J. M., Nery, J. R., … Ecker, J. R. (2012). Widespread dynamic DNA methylation in response to biotic stress. Proceedings of the National Academy of Sciences of the United States of America, 109, E2183–E2191. https://doi.org/10.1073/pnas.1209329109
Dubin, M. J., Zhang, P., Meng, D., Remiereau, M. S., Osborne, E. J., Casale, P. F., … Nordborg, M. (2015) DNA methylation in Arabidopsis
Morran, L. T., Schmidt, O. G., Gelarden, I. A., Parrish, R. C., & Lively, C. M. (2011). Running with the Red Queen: Host-parasite coevolution selects for biparental sex. Science, 333, 216–218. https://doi.org/10.1126/science.1206360

Nakamura, S., & Hosaka, K. (2010). DNA methylation in diploid inbred lines of potatoes and its possible role in the regulation of heterosis. Theoretical and Applied Genetics, 120, 205–214. https://doi.org/10.1007/s00122-009-1058-6

Negri, I., Franchini, A., Gonella, E., Daffonchio, D., Mazzoglio, P. J., Nakamura, S., & Hosaka, K. (2010). Unravelling the Wolbachia evolutionary role: The reprogramming of the host genomic imprinting. Proceedings of the Royal Society B: Biological Sciences, 276, 2485–2491. https://doi.org/10.1098/rspb.2009.0324

Neiman, M., Jokela, J., & Lively, C. M. (2005). Variation in asexual lineage age in Potamopyrgus antipodarum, a New Zealand snail. Evolution, 59, 1945–1952. https://doi.org/10.1111/j.0014-3820.2005.tb01064.x

Norouzitallab, P., Baruah, K., Vandegeechtume, M., Van Steappen, G., Catania, F., Vanden Bussche, J., … Bossier, P. (2014). Environmental heat stress induces epigenetic transgenerational inheritance of robustness in parthenogenetic Artemia model. The FASEB Journal, 28, 3552–3563. https://doi.org/10.1096/fj.14-252049

Peakall, R., & Smouse, P. E. (2012). GenAlEx 6.5: Genetic analysis in Excel. Population genetic software for teaching and research—an update. Bioinformatics, 28, 2537–2539. https://doi.org/10.1093/bioinformatics/bts460

Peng, H., & Zhang, J. (2009). Plant genomic DNA methylation in response to stresses: Potential applications and challenges in plant breeding. Progress in Natural Science, 19, 1037–1045. https://doi.org/10.1016/j.pnsc.2008.10.014

Perez-Figueroa, A. (2013). msap: A tool for the statistical analysis of methylation-sensitive amplified polymorphism data. Molecular Ecology Resources, 13, 522–527. https://doi.org/10.1111/1755-0998.12064

Pérez-Figueroa, A., García-Pereira, M., Saura, M., Rolán-Alvarez, E., & Caballero, A. (2010). Comparing three different methods to detect selective loci using dominant markers. Journal of Evolutionary Biology, 23, 2267–2276. https://doi.org/10.1111/j.1420-9101.2010.02093.x

Press, C. M., & Evensen, O. (1999). The morphology of the immune system in teleost fishes. Fish & Shellfish Immunology, 9, 309–318. https://doi.org/10.1006/fsim.1998.0181

Revelle, W. (2019). Package ‘psych’: The R Project for Statistical Computing. Available at: https://cran.r-project.org/web/packages/psych/psych.pdf

Richards, C. L., Schrey, A. W., Coon, C. A., Grispo, M. T., Awad, M., Imboma, T., McCoy, E. D., … Martin, L. B. (2012). Epigenetic variation may compensate for decreased genetic variation with introductions: A case study using house sparrows (Passer domesticus) on two continents. Genetics Research International, 2012, 1–7. https://doi.org/10.1155/2012/979751

Slowinski, S. P., Morran, L. T., Parrish, R. C., Cui, E. R., Bhattacharya, A., Lively, C. M., & Phillips, P. C. (2016). Coevolutionary interactions with parasites constrain the spread of self-fertilization into outcrossing host populations. Evolution, 70, 2632–2639. https://doi.org/10.1111/evol.13048

Smith, T. A., Martin, M. D., Nguyen, M., & Mendelson, T. C. (2016). Epigenetic divergence as a potential first step in dater speciation. Molecular Ecology, 25, 1883–1894. https://doi.org/10.1111/mec.13561

Steets, J. A., Wolf, D. E., Auld, J. R., & Ashman, T. L. (2007). The role of natural enemies in the expression and evolution of mixed mating in hermaphroditic plants and animals. Evolution, 61, 2043–2055. https://doi.org/10.1111/j.1558-5646.2007.00184.x

Tatarakos, A., Earley, R. L., Perlman, B. M., Taylor, D. S., Turner, B. J., & Avise, J. C. (2015). Genetic subdivision and variation in selfing rates among central american populations of the mangrove rivulus, Kryptolebias marmoratus. The Journal of Heredity, 106, 276–284. https://doi.org/10.1093/jhered/esv013

Tatarakos, A., Earley, R. L., Taylor, D. S., & Avise, J. C. (2012). Microevolutionary distribution of isogenicity in a self-fertilizing fish (Kryptolebias marmoratus) in the Florida Keys. Integrative and Comparative Biology, 52, 743–752. https://doi.org/10.1093/icb/icss07

Tatarakos, A., Gao, H., Mackiewicz, M., Taylor, D. S., Turner, B. J., & Avise, J. C. (2007). Strong population structure despite evidence of recent migration in a selfing hermaphroditic vertebrate, the mangrove killifish (Kryptolebias marmoratus). Molecular Ecology, 16, 2701–2711. https://doi.org/10.1111/j.1365-294X.2007.03349.x

Tatarakos, A., Lima, S. M. Q., Earley, R. L., Berbel-Filho, W. M., Vermeulen, F. B. M., Taylor, D. S., … Avise, J. C. (2017). Deep and concordant subdivisions in the self-fertilizing mangrove killifishes (Kryptolebias) revealed by nuclear and mtDNA markers. Biological Journal of the Linnean Society, 122, 558–578. https://doi.org/10.1093/biolinnean/btx103

Tatarakos, A., Lima, S. M. Q., Taylor, D. S., & Avise, J. C. (2009). Long-term retention of self-fertilization in a fish clade. Proceedings of the National Academy of Sciences of the United States of America, 106, 14456–14459. https://doi.org/10.1073/pnas.0907852106

Teh, A. L., Pan, H., Chen, L., Ong, M.-L., Dogra, S., Wong, J., … Holbrook, J. D. (2014). The effect of genotype and in utero environment on inter-individual variation in neonate DNA methylomes. Genome Research, 24, 1064–1074. https://doi.org/10.1101/gr.171439.113

Vähä, J. P., & Primmer, C. R. (2006). Efficiency of model-based Bayesian methods for detecting hybrid individuals under different hybridization scenarios and with different numbers of loci. Molecular Ecology, 15, 63–72. https://doi.org/10.1111/j.1365-294X.2005.02773.x

Van Valen, L. (1973). A new evolutionary law. Evolutionary Theory, 1, 1–30.

Venney, C. J., Johansson, M. L., & Heath, D. D. (2016). Inbreeding effects on gene-specific DNA methylation among tissues of Chinook salmon. Molecular Ecology, 25, 4521–4533. https://doi.org/10.1111/mec.13777

Vergeer, P., Wagemaker, N., & Ouborg, N. J. (2012). Evidence for an epigenetic role in inbreeding depression. Biology Letters, 8, 798–801. https://doi.org/10.1098/rsbl.2012.0494

Verhoeven, K. J. F., Jansen, J. J., van Dijk, P. J., & Biere, A. (2010). Stress-induced DNA methylation changes and their heritability in aseexual dandelions. New Phytologist, 185, 1108–1118. https://doi.org/10.1111/j.1469-8137.2009.03121.x
Verhoeven, K. J. F., & Preite, V. (2014). Epigenetic variation in asexually reproducing organisms. *Evolution*, 68, 644–655. https://doi.org/10.1111/evo.12320

Verhoeven, K. J. F., Vonholdt, B. M., & Sork, V. L. (2016). Epigenetics in ecology and evolution: What we know and what we need to know. *Molecular Ecology*, 25, 1631-1638. https://doi.org/10.1111/mec.13617

Wenzel, M. A., & Piertney, S. B. (2014). Fine-scale population epigenetic structure in relation to gastrointestinal parasite load in red grouse (*Lagopus lagopus scotica*). *Molecular Ecology*, 23, 4256–4273. https://doi.org/10.1111/mec.12833

Yu, A., Lepere, G., Jay, F., Wang, J., Bapaume, L., Wang, Y., … Navarro, L. (2013). Dynamics and biological relevance of DNA demethylation in Arabidopsis antibacterial defense. *Proceedings of the National Academy of Sciences of the United States of America*, 110, 2389–2394. https://doi.org/10.1073/pnas.1211757110

Zhang, Y. Y., Zhang, D. Y., & Barrett, S. C. H. (2010). Genetic uniformity characterizes the invasive spread of water hyacinth (*Eichhornia crassipes*), a clonal aquatic plant. *Molecular Ecology*, 19, 1774–1786. https://doi.org/10.1111/j.1365-294X.2010.04609.x

**SUPPORTING INFORMATION**

Additional supporting information may be found online in the Supporting Information section at the end of the article.

---

**How to cite this article:** Berbel-Filho WM, Garcia de Leaniz C, Morán P, Cable J, Lima SMQ, Consuegra S. Local parasite pressures and host genotype modulate epigenetic diversity in a mixed-mating fish. *Ecol Evol*. 2019;9:8736–8748. https://doi.org/10.1002/ece3.5426