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Local ancestry inference provides insight into Tilapia breeding programmes

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Tilapia is one of the most commercially valuable species in aquaculture with over 5 million tonnes of Nile tilapia, *Oreochromis niloticus*, produced worldwide every year. It has become increasingly important to keep track of the inheritance of the selected traits under continuous improvement (e.g. growth rate, size at maturity or genetic gender), as selective breeding has also resulted in genes that can hitchhike as part of the process. The goal of this study was to generate a Local Ancestry Interence workflow that harnessed existing tilapia genotyping-by-sequencing studies, such as Double Digest RAD-seq derived Single-Nucleotide Polymorphism markers. We developed a workflow and implemented a suite of tools to resolve the local ancestry of each chromosomal locus based on reference panels of tilapia species of known origin. We used tilapia species, wild populations and breeding programmes to validate our methods. The precision of the pipeline was evaluated on the basis of its ability to identify the genetic makeup of samples of known ancestry. The easy and inexpensive application of local ancestry inference in breeding programmes will facilitate the monitoring of the genetic profile of individuals of interest, the tracking of the movement of genes from parents to offspring and the detection of hybrids and their origin.

Despite their prominent role in aquaculture, the volume of research involving tilapia is relatively low when compared to other fish species, like salmonids. As a result, existing research on ancestry tracing in tilapia is not abundant either. A fast and accurate method for tracing the hybridisation of admixed fish, i.e. fish of mixed ancestry, would uncover their composition, thus reconstructing their origins and even aiding identification of escapees and monitor introgression of native species. It would also help follow the movement of unwanted or unexpected traits alongside selected ones in a population, thus yielding useful information to produce more economically and environmentally favourable variants. Local Ancestry Inference (LAI) applications are more frequent in studies of dog breeds, as in the case of Alaskan sled dogs[1], in which tracing of ancestry in sprint and long-distance sled dogs contributed to the identification of the genomic regions that correlated with performance-enhancing traits. Similarly, in humans, such tools have been more widely used to analyse how past migration events have impacted existing populations[2] and to improve identification of ancestry-specific genetic susceptibility to disease in genome-wide association studies[3].

Due to the relative scarcity of research specific to tilapia, or even fish in general, most of the literature currently available on inference of local ancestry focuses on human applications[4]. In over 15 years, more than 20 new LAI methods for human applications have been introduced[5]. Less often, relevant literature can be found on other animals like insects[6] or, as already mentioned, dogs[7].

While it is still possible to apply processes and tools developed for other animals to tilapia, a major obstacle persists, which is the vast difference in the amount, quality and variety of genotyped individuals available to build a reference panel. In humans, genetic studies often benefit from thousands[8], if not tens of thousands, of individuals of certain descent, as well as publicly available data like that produced by the 1000 genomes project[9]. In tilapia, only hundreds of individuals are usually available, and it is much more difficult to accurately trace specific families, which limits the variety of the reference samples and negatively impacts the accuracy of phasing and LAI tools.

The selective breeding of tilapia revolves around the creation and maintenance of variants which would ideally display the most economically and environmentally favourable traits of their ancestors. Tilapia, and in particular

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Species	Location	Origin	Size	Abbrev.	Referenecs
O. andersonii	Zambia	Reference species	6	AND	Syaifudin et al.[11]
O. aureus	Israel	Reference species	10	AUR-I	Syaifudin et al.[11]
	Lake Manzala, Egypt	Reference species	5	AUR-E	Syaifudin et al.[11]
O. karongae	Lake Malawi	Reference species	4	KAR	Syaifudin et al.[11]
O. macrochir	Zambia	Reference species	4	MAC	Syaifudin et al.[11]
O. mossambicus	South Africa	Reference species	13	MOS-A	Syaifudin et al.[11] and This study
	Zimbabwe	Reference species	9	MOS-Z	Syaifudin et al.[11] and This study
O. n. cancellatus	Ethiopia, Lake Hora	Reference species	14	CAN-H	Syaifudin et al.[11]
	Ethiopia, Lake Hora	Reference species	11	CAN-K	Syaifudin et al.[11]
O. n. filoa	Ethiopia, Lake Metahara	Reference species	8	CAN-M	Syaifudin et al.[11]
O. niloticus	Ghana, Lake Volta, Kpandu	Reference species	12	NIL-K	Syaifudin et al.[11]
	Ghana, Lake Volta, Nyinuto	Reference species	12	NIL-N	Syaifudin et al.[11]
	Lake Manzala, Egypt	Reference species	22	NIL-E	Syaifudin et al.[11] and This study
O. u. hornorum	Tanzania	Reference species	5	HOR	Syaifudin et al.[11]
S. galilaeus	Israel	Reference species	5	GAL	Syaifudin et al.[11]
S. melanotheron	Ghana	Reference species	4	MEL	Syaifudin et al.[11]
T. zillii	Ghana	Reference species	5	ZIL-G	Syaifudin et al.[11]
	Lake Manzala, Egypt	Reference species	5	ZIL-E	Syaifudin et al.[11]
Breeding programme	Colima, Mexico	Breeding programme	17	BRE-C	This study
	Morelos, Mexico	Breeding programme	18	BRE-M	This study
	Veracruz, Mexico	Breeding programme	18	BRE-V	This study
	Mexico	Breeding programme	18	BRE-L	This study
	Malaysia	GIFT programme	50	GIFT	Taslima et al.[13]

Table 1. Tilapia species and populations. Species, geographical location, origin, size and abbreviation of the 23 sample populations.

Nile tilapia (*Oreochromis niloticus*), are highly common among breeding programmes due to their relatively short reproduction cycle, hardiness, and resistance to disease and parasites[10]. It has become increasingly necessary to track the inheritance of selected traits under continuous improvement, as selective breeding may also result in genes to hitchhike along in the process. Implementation of LAI in breeding programmes allows the monitoring of the genetic makeup architecture of each individual, the tracking the genes inheritance from parents to offspring, and this ensures that only loci of interest are selected by the breeding programmes.

The goal of this project was to generate a LAI workflow that harnessed existing tilapia genotyping-by-sequencing studies[11–13], such as Double Digest RAD-seq (ddRAD) derived Single-Nucleotide Polymorphism (SNP) markers[14,15]. This provided an insight into breeding programmes with a more in-depth look at the genetic makeup of admixed individuals, significantly contributing to the identification of hybrids, and the development of new variants for aquaculture. We resolved the local ancestry of admixed individuals successfully and in detail, and the workflow was applied to the samples sourced from breeding programmes. We implemented a fast and accurate pipeline providing useful insights for breeding programmes of both tilapia and other animals, whether these are aimed at maintaining specific broodstocks or producing new variants.

Results

ddRAD library sequencing. High throughput sequencing of the animal from the four breeding programmes and additional individuals (93 individuals, Supplementary Table S1 online) produced 34,091,027 paired-end reads in total. After the filtering the reads, 82.5% of the total reads were retained (28,113,599 paired-end reads). The new reads as well as the published reads (275 samples; Supplementary Table S2 online) were mapped against the *O. niloticus* genome assembly (NCBI Assembly accession GCA_001858045.3). A total of 19,041 bi-allelic SNPs was extracted with a minor allele frequency (MAF) of at least 0.01, no deviation from the expected Mendelian segregation ($P > 0.01$) and common to at least 4 populations and 50% of their individuals (Table 1 and Supplementary Data S3 online).

Population structures. A Multidimensional scaling (MDS) analysis of identity by state (IBS) was utilised to separate the individuals into clusters based on their genetic distance[16]. This process grouped individuals of same origin together, while positioning the hybrids between the populations which more heavily contributed to their genome (Fig. 1).

The *O. n. filoa* (CAN-M) individual originating from Lake Metahara[17] was grouped with *O. n. cancellatus* (CAN-H/K). As their similarity has prompted a proposition for a re-classification of *O. n. cancellatus* as *O. cancellatus*, with two sub-populations, *O. c. cancellatus* and *O. c. filoa*[18], these samples were grouped with the remaining *O. n. cancellatus* populations due to the species being virtually indistinguishable.

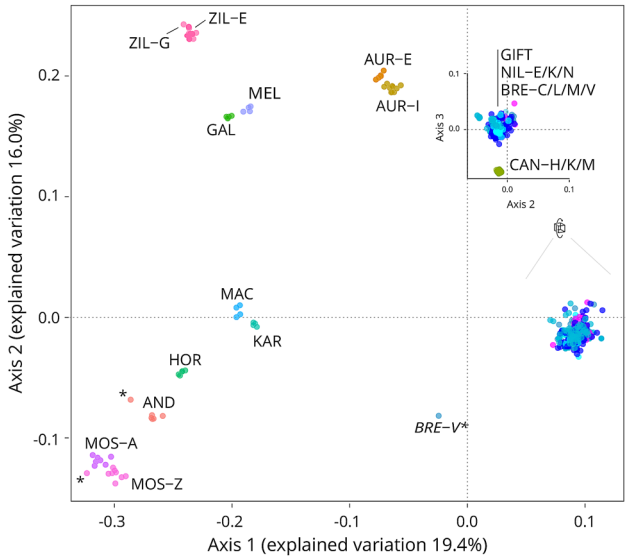

Figure 1. Multidimensional scaling analysis of identity by state results of the full dataset. First, second and third component explained 19.4%, 16.0% and 7.8% of the variation found. The inset on the top right, project the result on the second and third axes in order to distinguish the position of CAN-H/K and M compared to the NIL population and related populations (GIFT and BRE-C/L/M and V). Abbreviations included as listed in Table 1. *Indicate samples suspected to be hybrids (MOS-Z and AND) or outlier (BRE-V).

Most populations were clearly resolved (Fig. 1, Table 1 for abbreviations), with the exception of the breeding programmes populations (BRE-C/L/M/V and Genetically Improved Farmed Tilapia (GIFT) programmes), which remained tightly clustered together with NIL-E/K/N (*O. niloticus* populations, their species of expected origin). For most populations, species-specific grouping was representative of the genetic closeness of their samples, e.g. AUR-E and AUR-I (*O. aureus*), or the three CAN (*O. n. cancellatus*) sub-populations: while still distinguishable, these populations of different origin were clustered under one (species) group. The same could be said for *O. niloticus*: the NIL-E, NIL-K and NIL-N sub-populations were grouped under NIL. Multidimensional Scaling Analysis also highlighted the presence of outliers, especially in the form of one MOZ-Z (*O. mossambicus*, Zambia), one AND (*O. andersonii*) and one BRE-V individual (Fig. 1, marked by a *). BRE-V disparate positioning was found to be due to the high incidence of missing genotypes in some individuals, rather than due to sample impurity. Finally, the genetic closeness of some species was noted, especially of some of those only represented by a single small population (KAR and MAC, or GAL and MEL), and was expected to cause ambiguities when trying to resolve the ancestry of individual samples.

Ancestry inference. Before undergoing ancestry inference, these genotypic data were phased with BEAGLE[19]. Phasing is required to improve ancestry recognition, as separating the paternal and maternal contributions allows to infer the origin of each separately, since they could belong to different species or populations. Once these genotypic data are phased, RFMix separates each chromosome into a series of equally-sized windows, and the likelihood of each window belonging to each of the reference populations is calculated[20]. For each one, a random forest is trained to recognise the ancestry based on the reference panel: each tree of the random forest infers a putative ancestry, and a sum of all the *votes* determines the probabilities of that window originating from each possible ancestral population.

The inference accuracy using only the 155 references samples from 10 species was optimised to minimise the fragmentation, while maximising recognition of the reference samples (Fig. 2). The final combination featured 500 BEAGLE iterations, combined with 50 EM iterations and a SNPs window size of 7.

Digital chromosome painting. Using the reference samples as a training set, the breeding programme individuals were analysed for LAI. In contrast to the reference population, all of the individual samples exhibited a relatively high level of fragmentation (5 to 30%). As expected, the main contributor of the genome composition was *O. niloticus*, with variable contribution of *O. aureus* and *O. mossambicus* (Fig. 3). Several individuals showed a different composition (Fig. 2, samples marked with *).

Discussion

Regardless of the species, LAI studies follow similar steps. First, a large number of markers are gathered from populations of known descent to build a reference panel[21]. The genotypes then undergo phasing to reverse crossing-overs, separating the contributions of the two parents[22]. The reference panel is then used to train the LAI model[21,23]. This also includes a "smoothing" algorithm, which improve the results by solving phasing errors, in case the maternal and paternal contributions have been swapped at certain loci, as well as genotyping errors,

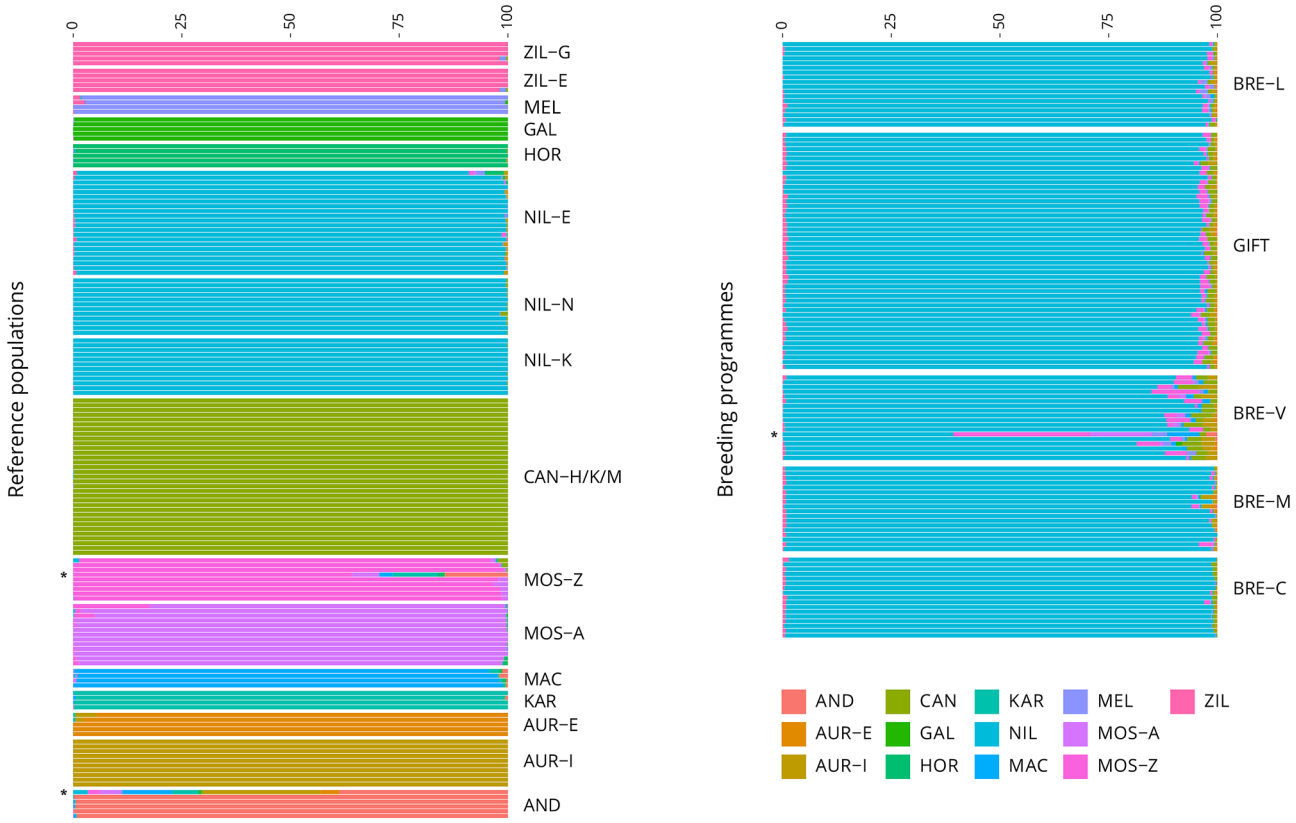

Figure 2. Global ancestry contribution. For each sample the predicted global ancestry contribution is reported. The reference sample global ancestry contributions were assigned based on a training set not including the breeding programme samples. * Indicate samples suspected to be hybrids (MOS-Z and AND) or outlier on the Multidimensional Scaling Analysis (BRE-V). Abbreviations included as listed in Table 1.

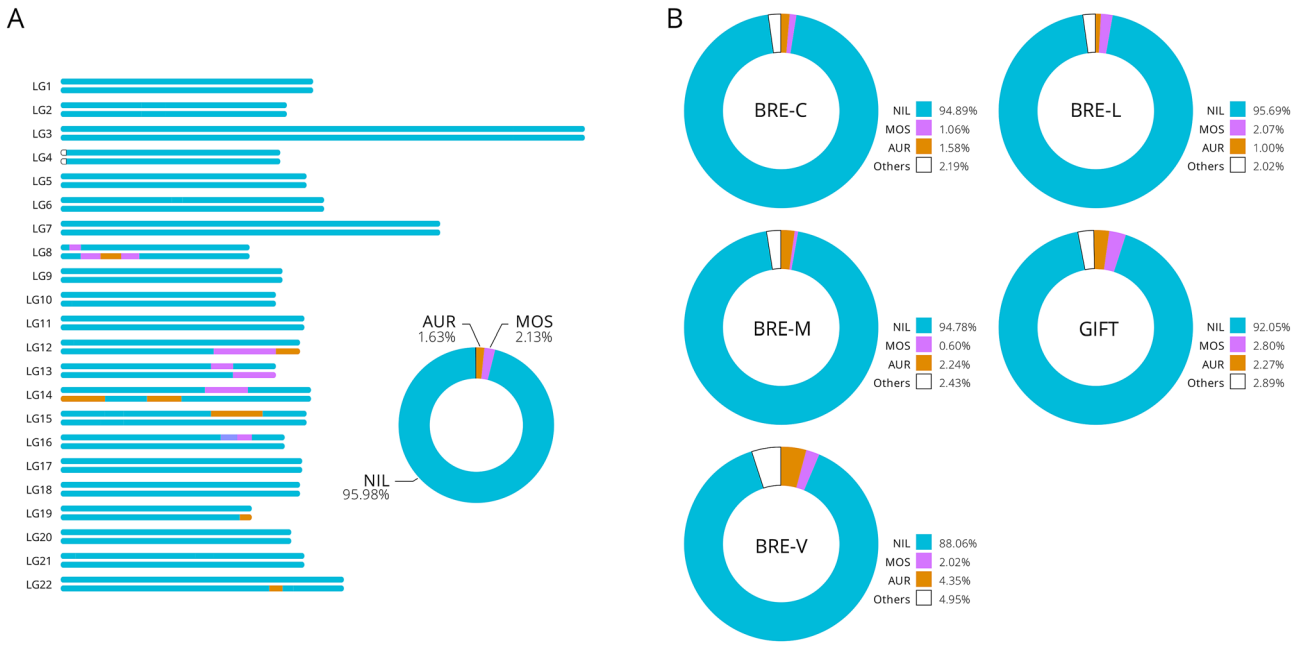

Figure 3. Local and Global ancestry of breeding programmes. (**A**) Local ancestry karyograms for the individual GIFT_12 (GIFT Breeding programme) with ancestry proportions similar to the estimated population averages. The *O. aureus* (AUR), *O. mossambicus* (MOS) and *O. niloticus* (NIL) haploid genomes present a complex mosaic of ancestry tracts across the genome, reflecting its demographic history; (**B**) Median ancestry proportions for breeding programmes, based on fractions of the chromosome length. Abbreviations included as listed in Table 1. Other, includes a low proportion of introgression that is more likely to be a consequence of a prediction ambiguity and noise rather than real introgression.

i.e. mistakes in the genome sequencing process that can be identified by their dissimilarity with the rest of the genome. Finally, the results are displayed graphically, on the chromosome[3], or by probability representations[24].

Intergenetic tilapia hybrids such as *Sarotherodon melanotheron* × *O. niloticus* as not uncommon and have been used in aquaculture to produce highly saline-tolerant hybrid[25]. Closer species crosses have commonly observed in the wild or feral populations. The ability of the methodology to correctly identify ancestries remains directly dependent on the "purity" and closeness of the reference samples, and on the quality of their genotyping: in fact, some species in this study could not be accurately recognised due to a high incidence of missing genotypes found in their samples (e.g. AND), the small sample number available (e.g. GAL or MEL), or because of their genetic closeness with other populations of small numbers (e.g. KAR and MAC), which prevented the algorithm from distinguishing the species appropriately.

While missing genotypes can be easily detected by examining the dataset, the same cannot be said for a lack of purity of the samples. Once the individuals are sampled and only sequenced SNPs remain from them, it is difficult to verify whether samples truly belong to the species and population they may be claimed to or if they have been mislabelled. Therefore, some of the uncertainty in the re-assignment of known ancestries may derive from unknown levels of admixture in supposedly pure samples. However, the ddRAD markers did support purity of species by tight species clustering.

O. n. cancellatus, also known as *Tilapia cancellata*, has been assimilated to *O. niloticus* since the sub-species was identified[26], and is considered to be synonymous. The same is true for *O. n. filoa*. In this study these sample grouped with *O. n. cancellatus*. The results produced by this project, however, draw a clear distinction between NIL and CAN, as the multidimensional scaling analysis indicates that although genetically close, the groups do not overlap (Fig. 1). If samples of *Tilapia cancellata* were considered to be part of *O. niloticus*, and the two were mixed to form a supposedly pure broodstock, *O. niloticus* would feature not only NIL, but also CAN markers, as well as contributions from all the other species that are considered synonymous to the species. This would explain the presence of CAN contamination in the hybrids, while also justifying the difference in accuracy of recognition of CAN versus NIL samples.

Similarly, further investigation of the relationship between *O. niloticus* and *O. aureus* revealed that the significant incidence of AUR contribution in the NIL samples as well as the breeding hybrids was not an isolated occurrence of this project. Different studies[17,27] observed that the two species are likely to share a common ancestral mitochondrial haplotype: the presence of some degree of hybridisation between these two species would explain why, even though AUR is genetically distant from all other populations, the AUR species has such a significant presence in the ancestry assignment of NIL and NIL-derived samples, while also being absent from all other species.

Regarding the quality of the breeding hybrids, the analysis of their ancestry showed that, on average, more than 90% of their genome derived from *O. niloticus*, with only minor contributions from other species. These results showed that the pipeline is capable of confidently recognising the ancestry of admixed individuals, as the hybrids analysed were indeed descendants of *O. niloticus*. Fish from the GIFT breeding programme have already been shown as having a small contribution from *O. mossambisus* genome and *O. aureus*[28–30]. The actual extent depends on the methodology used, but it ranges from 1 to 7% for *O. mossambisus* and 0.5% to 3% for *O. aureus*. In this study, which used a genome wide approach, we identified the contribution to be on average 3% and 2% for *O. mossambisus* and *O. aureus* respectively.

Hybridisation is common and affects original phenotypes in most of the areas. However, it lacks the suitable method to identify the individual derived from hybrid and introgression in a simple and inexpensive way. We report, the new pipeline which could be used in further to evaluate the most the marker-based studies without further expensive experimental sampling or sequencing. This methodology based on ddRAD SNP markers has shown itself capable of identifying the contribution of multiple ancestral populations to the genome of admixed individuals, both from a local and global perspective. If provided with a large body of fully genotyped populations of known origin, the results produced by this pipeline would contribute to a more informed breeding process for the creation and maintenance of tilapia variants.

Methods

Biological materials. Fin samples were collected from a total of 71 individuals from four breeding programmes located in Mexico: Colima (BRE-C), Morelos (BRE-M) and Veracruz (BRE-V) broodstock; descendants from the Institute of Aquaculture fish, over 15 years ago. BRE-L, YY fish were obtained from a stock originating in Costa Rica. An additional 6 *O. mossambicus* (3 MOS-A and 3 MOS-Z) and 16 *O. niloticus* (NIL) reference samples were included. Samples were stored in 95% ethanol at − 20 °C until required. Details of the samples and origins are listed in Table 1. Attempts were made to balance the sex ratios (Supplementary Table S1 online) in order to minimise any potential bias due to sex-specific regions of the genome.

Genomic DNA extraction. Purified DNA was extracted by a modified salt precipitation method[30]. Small pieces of fin tissue were digested in 300 µL SSTNE lysis solution (0.3 M NaCl, 50 mM Tris base, 0.2 mM EDTA pH 8.0, 0.2 mM EGTA, 0.5 mM spermidine, 0.25 mM spermine and 0.1% SDS) containing 1.5 µL Proteinase K (10 mg/mL) at 55 °C overnight. Lysed samples were treated with 5 µL RNaseA (2 mg/mL) at 37 °C for 1 h and the supernatant centrifuged twice at 21,000×*g* after precipitation with 180 µL 5 M NaCl on ice. The resulting DNA was precipitated in an equal volume of isopropanol, washed twice in 70% ethanol and dissolved in TE buffer (10 mM Tris, 1 mM EDTA pH 8.0) until DNA quantification. The quantity and quality of DNA were assessed by measurement on a Nanodrop spectrophotometer (Labtech International Ltd, UK) and by agarose gel electrophoresis. Standardised dilutions of 8 ng/µL DNA for each sample were prepared in 5 mM Tris buffer pH 8.0 according to fluorimetry values.

Double Digest RAD library preparation and sequencing. Two libraries were constructed (Supplementary Table S1 online) following the ddRAD library preparation protocol with slight modifications[11]. Briefly, for each library, individual DNA samples (36 ng–4.5 µL) were simultaneously digested with two high fidelity restriction enzymes (New England Biolabs, NEB, UK): *Sbf*I (CCTGCA|GG recognition site), and *Sph*I (GCATG|C recognition site). Digestions were incubated for 90 min at 37 °C, using 0.72 U of each enzyme in 1× CutSmart Buffer (NEB) and in a 9 µL reaction volume. The reactions were then cooled to 22 °C, 4.5 µL of a pre-made barcode/adaptor mix was added to each digested DNA sample and incubated at 22 °C for 10 min. The adaptor mix included individual-specific barcoded combinations of P1 (*Sbf*I-compatible) and P2 (*Sph*I-compatible) adaptors at 6 nM and 72 nM concentrations respectively, in 1× reaction buffer 2 (NEB). The adaptors included an inline five- or seven-base barcode for sample identification. Ligation was performed over 2.5 h at 22 °C by addition of a further 4.5 µL of a ligation mix including 4 mM rATP (Promega, UK), and 2000 cohesive-end units of T4 ligase (NEB) per µg DNA in 1× CutSmart buffer. Samples for each library were combined into a single pool. The pooled libraries were column-purified (MinElute PCR Purification Kit, Qiagen, UK), and eluted in 60 µL EB buffer (Qiagen, UK). Size-selection of fragments, ranging from 320 to 590 bp, was performed by agarose gel separation. Following gel purification (MinElute Gel Extraction Kit, Qiagen, UK), the eluted size-selected template DNA (65 µL in EB buffer) was PCR amplified (11–12 cycles PCR dependent on library; 32 separate 12.5 µL reactions, each with 1.25 µL template DNA) using a high fidelity Taq polymerase (Q5 Hot Start High-Fidelity DNA Polymerase, NEB). The PCR reactions were combined (400 µL total), and column-purified (MinElute PCR Purification Kit). The c. 50 µL eluate, in EB buffer, was then subjected to a further size-selection clean up using an equal volume of AMPure magnetic beads (Perkin-Elmer), to maximise removal of small fragments (less than c. 200 bp). Each final library was eluted in 15 µL EB buffer, QUBIT quantified and diluted to 10 nM stocks and sequenced in house on a separate Illumina MiSeq run (v2 chemistry, 300 cycle kit, 150 base paired-end reads).

Data origins. A total of 10 different species[11,12], along with individuals sourced from breeding programmes, were used to produce ddRAD markers[31] following the same protocol: restriction enzymes set (*Sbf*I and *Sph*I), size selection (320 bp to 590 bp) and comparable sequencing platforms (150 nucleotide paired-ends), rendering their results compatible. Efforts were made to use populations with known histories, an absence of hybridisation, and from multiple locations (Table 1). The *O. niloticus* samples consisted of three sub-species (*O. niloticus sensu stricto* and *O. n. filoa* and *O. n. cancellatus*); *O. aureus*, *O. mossambicus* and *Tilapia zillii* (Gervais: reclassification as *Coptodon zillii* proposed by Dunz and Schliewen[32]) comprised samples from two locations each, while *O. karongae* (Trewavas), *O. urolepis hornorum* (Norman), *O. andersonii*, *O. macrochir*, *Sarotherodon galilaeus* (Linnaeus) and *S. melanotheron* consisted of samples from one location each. As far as could be ascertained, each originated from a single wild population (in some cases then maintained and bred in captivity). Additionally, a ddRAD dataset from the popular Genetically Improved Farmed Tilapia (GIFT) breeding programme[13] and samples from four breeding programmes in Mexico were assessed (Table 1).

Dataset preparation. Reads of low quality (i.e., with an average quality score less than 20), lacking the restriction site or having ambiguous barcodes were discarded during the samples demultiplexing stage. Retained reads were aligned against the genomic assembly of the tilapia species *O. niloticus* (NCBI Assembly accession GCA_001858045.3) using bwa[33] and assembled using Stack[34]. Markers produced through ddRAD sequencing were collected from the 275 samples. All loci that were common to at least 4 populations and at least 50% of their individuals, a minor allele frequency over 0.05 and not deviating from the expected Mendelian segregation ($P > 0.01$) were retained, as the missing data could be inferred by imputation.

Ancestry inference. BEAGLE[19] was used for the phasing of genotypes. BEAGLE performs multiple phasing iterations per SNP. After the phasing was carried out and the model was fit, the data were analysed again to obtain new estimates that allowed a better refit of the model. RFMix[23] was used for LAI. To optimise the inference accuracy using only the 155 references samples from 10 species, the number of phasing iterations, number of expectation-maximisation (EM) iterations, and the chromosomal window size were varied, and their results were compared. The combination of parameters that produced the least amount of fragmentation in theoretically pure individuals was chosen as most suitable.

Multidimensional scaling analysis. R v3.5.2[35] was used to carry out Multidimensional Scaling Analysis on the dataset using the package R/SNPRelate v1.16.0[36] to calculate the Identity-By-State (IBS) proportion for each sample.

Digital chromosome painting. Inferred local ancestry data, produced by RFMix, were visualised using R for the distribution of the local probabilities, and a dedicated script rendered the final distribution as a painted karyotype for each sample. Full scripts and pipelines are available on GitHub at https://github.com/pseudogene/fish_pedigree.

Ethical approval. Animal handling and collection was conducted under the UK Home Office guidelines and regulations [Samples MOZ-A/Z and NIL] and the Michoacán de Ocampo authority (Mexico) guidelines and regulations [Samples from the Breeding programme; BRE-C/L/M/V]. The ethical approval for the study was obtained from the University of Stirling (UK) Ethical committee. The data analytics and bioinformatics were

assessed by the Institute of Aquaculture Ethical Review Committee and passed the University of Stirling Ethical Review Process.

Data availability

The raw sequencing reads generated for this article are available from the EBI/ENA, accession Numbers ERR4170942 and ERR4171069, project PRJEB38387.

Code availability

The versions, settings and parameters of the software used in this work are as follows: (1) **process_radtags.pl**: Stack version 2.3, parameters: -E phred33 -filter_illumina -s 20 -c -q -t 135 -inline_inline -renz_1 sbfI -renz_2 sphI; (2) **bwa**: version 0.7.17, default parameters; (3) **ref_map.pl**: Stack version 2.3, parameters: -unpaired; (4) **populations**: Stack version 2.3, parameters: -write-single-snp -p 4 -r 50 -R 50 -min-maf 0.05 -hwe -vcf; (5) **BEAGLE**: version 5.1, parameters: java -Xmx30688m -jar beagle.21Sep19.ec3.jar imp-nsteps=50 iterations=1000 window=0.7 overlap=0.07; (6) **RFMix**: version 2.03, parameters: -e 50.

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Author contributions

A.A. performed the analytic experiments, interpreted these data, wrote the initial draft of the manuscript; M.B. conceived the study, interpreted the data and wrote the final manuscript. K.T. provided the extra *O. mossambicus* and *O. niloticus* samples (NIL) and completed the ddRAD libraries. A.C.M. provided the samples from Mexico (BRE-C/M/V/L). K.L.B. completed the ddRAD-seq on the breeding programmes, interpreted the data and wrote the final manuscript. S.L.C.S. processed the breeding programmes samples. All authors read and approved the final manuscript.

Competing interests

The authors declare no competing interests.

Additional information

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