

1 **High-throughput single nucleotide polymorphism (SNP) discovery and validation through whole-  
2 genome resequencing of hundreds of individuals in Nile tilapia (*Oreochromis niloticus*)**

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## ABSTRACT

26 Nile Tilapia (*Oreochromis niloticus*) is the second most important farmed fish in the world and a  
27 sustainable source of protein for human consumption. Several genetic improvement programs have been  
28 established for this species in the world and so far, they are mainly based on conventional selection using  
29 genealogical and phenotypic information to estimate the genetic merit of breeders and make selection  
30 decisions. Genome-wide information can be exploited to efficiently incorporate traits that are difficult to  
31 measure in the breeding goal. Thus, SNPs are required to investigate phenotype–genotype associations  
32 and determine the genomic basis of economically important traits. We performed *de novo* SNP discovery  
33 in three different populations of farmed tilapias. A total of 29.9 million non-redundant SNPs were  
34 identified through Illumina (HiSeq 2500) whole-genome resequencing of 326 individual samples. After  
35 applying several filtering steps including removing SNP based on genotype and site quality, presence of  
36 Mendelian errors, and non unique position in the genome, a total of high quality 50,000 SNP were selected  
37 for validation purposes. These SNPs were highly informative in the three populations analyzed showing  
38 between 43,869 (94%) and 46,139 (99%) SNP in HWE; 37,843 (76%) and 45,171(90%) SNP with a MAF  
39 higher than 0.05 and; 43,450 (87%) and 46,570 (93%) SNPs with a MAF higher than 0.01. The final list  
40 of 50K SNPs will be very useful for the dissection of economically relevant traits, enhancing breeding  
41 programs through genomic selection as well as supporting genetic studies in farmed populations Nile  
42 tilapia using dense genome-wide information.

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44 *Keywords:* Single Nucleotide Polymorphism; *Oreochromis niloticus*; next-generation sequencing;  
45 Illumina; genomic selection

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## INTRODUCTION

50 The study of phenotype-genotype association, identification of the genomic basis of economically  
51 important traits and the implementation of genomic predictions in farmed fish require a considerable  
52 number of highly informative single nucleotide polymorphisms (SNP) that preferably segregate in  
53 multiple populations. Thus, the discovery and characterization of dense SNP panels will help to a better  
54 understanding of complex traits architecture and genome biology in farmed fish (Yáñez et al. 2015). From  
55 an animal breeding perspective, the use of a high number of SNPs markers to support tilapia genetic  
56 improvement programs has the potential to speed up genetic gains for traits which by their nature cannot  
57 be directly recorded in selection candidates e.g. carcass quality and disease resistance traits (Yáñez and  
58 Martinez 2010; Ødegård et al. 2014; Yáñez et al. 2014). Dense SNP panels can also allow the  
59 determination of genomic regions underlying selection and adaptation to different environmental  
60 conditions during the domestication process in farmed fish populations (López et al. 2015).

61 The discovery of SNP markers in aquaculture species of commercial interest has been widely  
62 spread due the availability of high quality reference genomes as it is the case of Atlantic salmon (*Salmo*  
63 *salar*) (Lien et al. 2016), rainbow trout (*Oncorhynchus mykiss*) (Berthelot et al. 2014) and Pacific oyster  
64 (*Crassostrea gigas*) (Zhang et al. 2012). These information has facilitated the development of dense SNP  
65 panels, being currently available for different species including Atlantic salmon (Houston et al. 2014;  
66 Yáñez et al. 2016), rainbow trout (Palti et al. 2015), channel catfish (*Ictalurus punctatus*) (Liu et al. 2014;  
67 Zeng et al. 2017) and Pacific oyster (Gutierrez et al. 2017; Qi et al. 2017). These genomic resources have  
68 been used to carry out several studies aimed at identifying the genetic architecture of economically  
69 relevant traits in fish by means of genome-wide association studies for traits such as growth (Gutierrez et  
70 al. 2015; Tsai et al. 2015; Yoshida et al. 2017; Neto et al. 2019), disease resistance (Correa et al. 2015;  
71 Tsai et al. 2016; Barría et al. 2018) and carcass quality (Gonzalez-Pena et al. 2016). These SNP panels  
72 have also been used to test different approaches for the implementation of genomic predictions in Atlantic

73 salmon (Ødegård et al. 2014; Bangera et al. 2017; Correa et al. 2017; Sae-Lim et al. 2017) and rainbow  
74 trout (Vallejo *et al.* 2016, 2017, 2018; Yoshida, *et al.* 2018a; Yoshida, *et al.* 2018b).

75 Nile tilapia (*Oreochromis niloticus*) is among the most important fresh-water species farmed  
76 worldwide. Several selective breeding programs have been established for this specie since 1990's,  
77 allowing to genetically improve important commercial traits and expand tilapia farming across the globe.  
78 To date, the most widespread improved tilapia strain is the Genetically Improved Farmed Tilapia (GIFT)  
79 (Webster and Lim 2006), being farmed in Latin America, Asia, and Africa (Gupta and Acosta 2004). It  
80 has been shown that the response to selection for growth rate reached up to a 15% per generation after six  
81 generations of selection (Ponzoni et al. 2011), demonstrating the feasibility to improve this trait by means  
82 of artificial selection. However, and despite the large number of genetic programs and the advantages of  
83 Nile tilapia farming (e.g. fast growth and high adaptability), there are scarce studies on the application of  
84 genomic technologies for mapping variants associated with desired traits and enhancing selection through  
85 the use of genomic predictions in comparison with other aquaculture species. Consequently, up to date,  
86 genetic improvement programs mainly rely on traditional pedigree-based breeding approaches, with only  
87 one published report on the development of genome resources to enhance selective breeding in a single  
88 Nile tilapia population (Joshi et al. 2018).

89 The objective of this study was to perform a large-scale *de novo* SNP discovery and using whole genome  
90 resequencing of hundreds of Nile tilapia individuals from three different farmed populations and  
91 characterize a medium-density (50K) SNP panel to be further used in the determination of the genetic  
92 basis of complex traits and genomic selection in this species.

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## 95 MATERIALS AND METHODS

### 96 **Populations**

97 The principal aim of the present study was to discover and characterize highly informative SNP variants  
98 in Nile tilapia farmed populations. Thus, we included animals from three different commercial breeding  
99 populations established in Latin America, originated from admixed stocks imported from Asia. We used  
100 59 samples from POP A breeding population (Brazil) and 126 and 141 samples from POP B and POP C  
101 breeding populations, respectively, both belonging to Aquacorporación Internacional (Costa Rica). The  
102 three breeding populations are directly or indirectly related to the GIFT (Genetically Improved Farmed  
103 Tilapia), which is the most spread Nile tilapia strain used for farming purposes worldwide. The GIFT  
104 strain was initially established in Philippines by the crosses between four farmed Asian strains originally  
105 from Israel, Singapore, Taiwan and Thailand and four wild strains from Egypt, Senegal, Kenya and Ghana.  
106 The POP A breeding population represents GIFT animals which were introduced to Brazil for  
107 multiplication and farming purposes in early 2000. The POP B breeding population is a mixture of the  
108 original Asian farmed populations from Israel, Singapore, Taiwan and Thailand present in the Philippines  
109 in the late 1980s, which give origin to the GIFT strain. The POP C breeding population represents a  
110 combination of genetic material from the best available stocks corresponding to GIFT (Generation 8) and  
111 two original African strains founding GIFT. The three populations have been genetically improved for  
112 growth rate for more than 10 generations in total, using genetic evaluations based on the best linear  
113 unbiased predictor.

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## 115 **Whole-genome resequencing**

116 Tissue samples from the 326 fish were obtained by partial fin-clipping of fish anesthetized using  
117 benzocaine. Tissue sampling was carried out in accordance with the commercial practice and norms held  
118 by the two companies, Aquacorporación Internacional and Aquamerica, which provided the samples.  
119 Genomic DNA was extracted from fin-clip samples using the DNeasy Blood & Tissue Kit (QIAGEN)

120 following manufacturer's protocol. Whole-genome resequencing was performed on each of the individuals  
121 multiplexing five bar-coded samples per lane of 150 bp paired-end in Illumina HiSeq-2500.

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123 **SNP discovery and annotation**

124 We used the assembly ASM185804v2 (GenBank accession GCF\_001858045.1) of the *O. niloticus* as a  
125 reference genome sequence. This assembly consists of 1,010 Mb of total sequence comprising 2,990  
126 contigs with a contig N50 of 3.09 kb. Sequences from all samples were evaluated using FASTQC (Wingett  
127 and Andrews 2018) to assess base quality and primer adapter contamination. Burrows-Wheeler Aligner  
128 (BWA-MEM) (Li and Durbin 2009) was used to map the reads of each sample to the reference genome.  
129 Briefly, BWA-MEM starts a local alignment between a fragment of the read to the reference genome and  
130 extends it until the read is completely mapped, if the read cannot be fully mapped is soft clipped or  
131 eventually discarded. To avoid invalid flags in further analysis, reads without a pair were discarded from  
132 the output using SAMtools (Li et al. 2009). In order to obtain a high quality BAM file all duplicated reads  
133 were masked as such using PICARD (<http://broadinstitute.github.io/picard>). For variant calling we used  
134 the standard protocol implemented in the Genome Analysis Tool (GATK) version 3.5.0. All high-quality  
135 BAM files for each sample obtained previously, were assessed at the SNP calling step and summarized  
136 into a single Genotyped Variant Calling Format file (VCF) containing all data. Each SNP was categorized  
137 as being either homozygous or heterozygous for the ALT allele (i.e., the non-REF allele). To call a sample  
138 homozygous for an ALT allele at a given site, the most common ALT allele variant confidence divided  
139 by the total sample reads (QD) must be at least 10 (QD > 10). This normalized ALT alleles in zones with  
140 high density depth and poor-quality calls. Only bi-allelic SNPs were pre-selected in posterior filters. The  
141 final VCF file was annotated using Variant Effect Predictor (VEP v92.1) in offline mode using the cached  
142 Orenil1.0 genome database and the gff file GCF\_001858045.1\_ASM185804v2\_genomic.gff.

143

144 **SNP filtering and validation**

145 Population genetics analyses and filtering described here, including Hardy-Weinberg Equilibrium (HWE),  
146 minor allele frequencies (MAF) and observed and expected heterozygosities ( $H_O$  and  $H_E$ , respectively),  
147 were carried out using *VCFTools* (Danecek et al. 2011) and *Plink* (Purcell et al. 2007). An initial common  
148 quality control (QC) for the three populations was performed using VCF tools software. The genotypes  
149 were filtered to remove indels and sequence alterations, markers with genotypes quality ( $GQ < 0.15$ ) and  
150 minimum quality ( $minQ < 40$ ). Furtherly, specific QC filters for each population were applied discarding  
151 SNPs with missing genotypes  $> 0.60$ , minor allele frequency (MAF)  $< 0.01$ , Hardy-Weinberg Equilibrium  
152 (HWE)  $p$ -value  $< 1e-06$ , Illumina score  $< 0.8$  (Table 1). SNPs were also filtered based on Mendelian error  
153 using genotypes from 8 trios (sire, dam and offspring) from POP B, in which markers with less than one  
154 Mendelian error were retained. In addition, SNP probes were aligned to the Nile tilapia reference genome  
155 to retain markers which have a unique position in the genome assembly (GenBank accession  
156 GCF\_001858045.1) generated by the University of Maryland and the University of Stirling (Conte et al.  
157 2017) using the following procedure: i) SNP probes of 121 bp were built using flanking SNPs sequences  
158 (60 bp upstream and 60 bp downstream of each SNP); and ii) each probe was aligned to the reference  
159 genome by means of BLASTN version 2.3.0+ (Madden 2002), using the following parameters: word size  
160 of 11 (-w) and minimum e-value of e-40 (e); iii) all hits were evaluated tolerating only 2 mismatches, and  
161 no gaps were allowed; and iv) probes having a unique location in the genome were retained. All this  
162 procedure was achieved using in-house Python scripts. Furtherly, SNPs with MAF  $> 0.05$  in the three  
163 commercial populations were prioritized. Finally, SNPs were selected so that they are as evenly distributed  
164 along the genome as possible. This was done by selecting SNPs from windows of equal size across various  
165 chromosomes of the genome using THIN  $< 9$  kb command. When selecting SNPs from windows, higher  
166 preference was given to common SNPs between population POP B and POP C.

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168 **Animals ethics approval**

169 DNA sampling was carried out in accordance with the commercial practice and norms by  
170 Aquacorporación Internacional and Aquamerica.

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173 **RESULTS**

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175 **SNP discovery**

176 Whole-genome resequencing of 326 fish yielded a mean of 79.6 (SD = 65.0) millions of raw reads per  
177 fish, with a minimum and maximum of 20.6 and 545.6 millions of raw reads per fish, respectively. Quality  
178 controlled reads were aligned to the Nile tilapia reference genome, and an average of 76.3 (SD = 64.6)  
179 million read per fish, with a minimum and maximum of 20.5 and 543.1 million reads per fish, respectively,  
180 could be confidently and uniquely mapped to a single position in the genome and these were used for SNP  
181 discovery. Thus, the mean coverage for each fish was 8.7x (SD = 8.9x), with a minimum and maximum  
182 of 2.1x and 65.7x coverage per fish, respectively. After the SNP discovery phase, approximately  
183 38,454,404 sequence variants were identified across the panel of 326 individuals. A total 29,956,401 non-  
184 redundant SNPs were identified across the panel of 326 fish, and 26,415,097 (88.17%) of these SNPs  
185 passed the genotypes quality (GQ<0.15) and minimum quality (minQ<40) filters (Table 1). After  
186 discarding 1,596 SNPs from mitochondria, specific QC filters were applied for each population separately,  
187 removing SNP based on missing genotypes >0.60, MAF <0.01, HWE P-value < 1e-06, at least one  
188 Mendelian error assessed in trios from POP B and non-unique position of SNP probes in the Nile tilapia  
189 genome. A total of 261,550; 887,072 and 461,645 SNP were retained after the filtering steps mentioned  
190 above for POP A, POP B and POP C, respectively. From all these high quality SNP variants, only 31,694  
191 were common between the three populations and 238,025 SNP variants were common between the two

192 high priority populations POP B and POP C. After applying THIN < 9 kb command in order to select SNP  
193 as evenly distributed along the genome as possible only 16,275 SNP were common between the three  
194 populations, which were used as the base. The gaps to have a mean of one SNP every 9 kb were filled  
195 with additional 33,769 SNP common between POP B and POP C to reach a total of 50,044. Out of these  
196 50,04 SNPs, 44 SNPs from short unplaced scaffolds were removed.

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## 198 **SNP distribution and annotation**

199 To determine the distribution of SNPs in the Nile tilapia genome, we identified their chromosome and  
200 position into the public GenBank accession assembly GCF\_001858045.1 produced by the University of  
201 Maryland and the University of Stirling (Conte et al., 2017). The SNPs cover 1.01 Gb of the total assembly  
202 length and averaged one SNP every 9 kb. A total of 47,349 SNP (94.70%) were located in chromosomes  
203 and 2,651 SNP were located into unplaced scaffolds. After SNP annotation, we found that most of the  
204 uniquely anchored SNPs were located in introns (57.81%). Further a total of 12.2%, 11.97%, 7.16% and  
205 0.63 were located downstream, upstream, intergenic and exon regions, respectively. The remaining SNPs  
206 were found in splice acceptor, splice donor, splice site, 3'UTR and 5'UTR regions. The Pearson  
207 correlation coefficient between the number of SNPs within each chromosome and total chromosome size  
208 in terms of Mb is  $r = 0.95$  (p-value  $< 2.24e^{-11}$ ). The relationship between the number of SNPs per  
209 chromosome and the total chromosome length in Mb is shown in Figure 1. Thus, the discovered SNP  
210 present an even distribution across the chromosomes on the Nile tilapia genome assembly.

211

## 212 **SNP validation and population segregation**

213 We also performed comparisons between different populations in terms of population genetic estimates  
214 using a 50K SNP validation panel. In this respect, the percentage of SNP segregating in HWE in all the  
215 populations was 99%, 98% and 94% of the 50K SNP validation panel for POP A, POP B and POP C,

216 respectively. Furthermore, these SNPs showed 76% and 87%, 89% and 93%, and 90% and 93% of MAF  
217 > 0.05 and MAF > 0.01 for POP A, POP B and POP C, respectively (Table 2). The distribution of MAF  
218 values across SNPs ranged from 0.04 to 0.50 with mean MAF value of  $0.24 \pm 0.12$  (Figure 2). The average  
219 observed and estimated heterozygosity ( $H_o$  and  $H_E$ ) was evaluated in each population (Table 2). Although  
220 the  $H_o$  values were very similar among populations, POP A and POP B expressed the lowest (0.20) and  
221 the highest (0.25)  $H_o$  values, respectively, suggesting that these populations are the least and the most  
222 genetically diverse populations in the present study. In the three populations,  $H_o$  diverged considerably  
223 from  $H_E$ , resulting in a heterozygote deficiency compared to HWE expectations.

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## DISCUSSION

226 The application of molecular markers into breeding programs has been widely spread along terrestrial and  
227 aquaculture species. Dense SNP panels have been shown to facilitate genome-scale studies by allowing  
228 the simultaneous evaluation of thousands of SNPs in commercially important fish species, such as Atlantic  
229 salmon (Houston et al. 2014; Yáñez et al. 2016) and rainbow trout (Palti et al. 2015). These markers have  
230 facilitated the discovery of genetic variants associated with important commercial traits and also the  
231 evaluation and implementation of genomic selection in aquaculture species (Correa et al. 2015;  
232 Palaiokostas et al. 2016; Bangera et al. 2017; Gutierrez et al. 2018; Vallejo et al. 2018). However, and  
233 despite Nile tilapia is widely produced in several countries, with the existence of more than 20 breeding  
234 programs (Neira 2010), there are still scarce studies aiming at the application of genome-wide SNP  
235 information for the identification of quantitative trait loci and the evaluation and practical implementation  
236 of genomic predictions in this species. The SNP discovery strategy used here allowed us to identify a large  
237 number of high quality SNPs that can reliably be genotyped across different populations of farmed Nile  
238 tilapia with a GIFT origin. The GIFT strain is the most spread Nile tilapia strain used for farming purposes  
239 worldwide (Ponzoni et al. 2011). The results from the segregation of SNPs between different populations

240 indicate that the molecular markers identified in the present study would be useful for genetic studies  
241 across populations, although the performance of this set of markers would slightly decrease when used in  
242 POP A. This is most likely due to the genetic differentiation between populations associated with their  
243 distinct origin (founder effect) and independent genetic selection by more than ten generations. The  
244 emphasis placed on including SNPs segregating in POP B and POP C may have caused ascertainment  
245 bias, which most likely contributed to the lower diversity observed in the POP A. In addition, there is  
246 difference in the number of SNPs with MAFs higher than 0.05 and 0.01 for POP A compared to POP B  
247 and C. Therefore, these considerations must be taken into account when using the current SNP panel in  
248 farmed Nile tilapia populations with different origins and even on wild populations.  
249 A recent study has shown the development of a 58K SNP array for Nile tilapia by means of SNP discovery  
250 performed using whole-genome resequencing data of 32 fish from one commercial population (Joshi *et*  
251 *al.*, 2018). In this previous study, 40,549 (69.35%) out of 58,466 SNPs were retained after filtering by  
252 MAF  $\leq 0.05$ . In our study, between 37,843 (75.68%) and 45,171 (90.34) out of the 50K SNP validation  
253 panel were retained after filtering by HWE and MAF  $\leq 0.05$ , indicating a better proportion of SNP  
254 validated and a moderate variation (~15%) of availability of SNPs, depending on the target population.  
255 The latter is most likely due to ascertainment bias in SNP discovery and selection and it has to be taken  
256 into account in further applications of this SNP panel in populations with different origins. When  
257 comparing the SNP list from the 50K SNP validation panel against the 58K SNP array developed by Joshi  
258 *et al.* (2018), by means of aligning SNP probes, we found that 100% of the SNPs were exclusive to each  
259 SNP panel. The high proportion of SNPs exclusive to each of the two SNP panels can be mainly explained  
260 by the different genetic background of populations and design of the whole-genome resequencing  
261 experiments used for SNP discovery. The 50K SNP validation panel presented here was produced using  
262 whole-genome resequencing of 326 fish from three independent populations, which allowed us to have an  
263 initial list of 29.9 million putative SNPs, which was almost a three times larger initial set when compared

264 against the previous study from Joshi et al. (2018), in which 32 fish from a single population were whole-  
265 genome resequenced, generating 10.5 million putative SNPs for further filtering steps. More importantly,  
266 the results presented here indicate that currently available Nile tilapia SNP panels can be considered more  
267 as being highly complementary than redundant in terms of the variants represented.  
268 The SNP panel presented here provides an excellent resource for the development of genome-scale studies  
269 of biologically and economically important traits. For instance, a recent genome-wide association study  
270 using a subset 2.4 million SNPs derived from the 29.9 million SNPs available from the present study,  
271 confirmed the anti-Müllerian hormone as a major gene associated with sex determination in different  
272 populations of farmed Nile tilapia (Caceres et al. 2019). This information could assist future strategies  
273 aiming at generating monosex (all-male) Nile tilapia populations for farming purposes without using  
274 hormones, to better exploit the sexual dimorphism present in the species, in which male individuals growth  
275 faster than females (Baroiller and D'Cotta 2001; Alcantar et al. 2014). In addition, the SNP panel  
276 developed in the present study will also allow the practical implementation of genomic predictions in Nile  
277 tilapia selective breeding programs, as it has been reported in a recent study in which an increase in  
278 accuracy of EBVs has been demonstrated through the incorporation of genomic information into genetic  
279 evaluations for fillet traits (Yoshida et al. 2019a). Finally, the SNP resources presented here will also allow  
280 other kind of population genetic studies in farmed populations of Nile tilapia using dense genome-wide  
281 information, as for example, has been recently done by the determination of the genetic structure and  
282 linkage disequilibrium in farmed populations using dense SNP genotypes (Yoshida et al. 2019b).

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## CONCLUSIONS

285 This paper describes the simultaneous discovery and validation of SNP markers in Nile tilapia through the  
286 use of whole-genome resequencing of hundreds of animals. The SNPs identified here will provide an  
287 opportunity for the dissection traits of biological and economic importance, such as growth, carcass quality

288 and disease resistance traits, through the application in genome-scale studies. Furthermore, it will allow  
289 increasing the response to selection for these traits by means of genomic selection in breeding programs.  
290 We believe that downstream applications of this important genomic platform will help to enhance Nile  
291 tilapia production by making it more efficient and sustainable.

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304

305 Author Contributions

306 J.M.Y. conceived and designed the study, contributed to the analysis and drafted the manuscript. G.Y.  
307 contributed to analysis and writing. A.B. drafted the first version of the manuscript. G.C., M.E.L. and A.J.  
308 participated in data collection, purification and management of the samples for sequencing and  
309 genotyping. R.P., D.D., D.T., and A.M. performed the bioinformatics analysis and contributed to writing.  
310 J.P.L. participated in the design of the study and writing. JS and DS contributed to samples collection. All  
311 authors have reviewed and approved the manuscript.

312

313 Conflict of Interests

314 Two commercial organizations (Aquainnovo and Illumina) were involved in the SNP identification and  
315 preparation of the manuscript. However, this does not alter public accessibility to data from the SNP data  
316 presented in this study. JPL was employed by Benchmark Genetics Chile during the course of the study.

317

318 Data Availability

319 The sequence data used for SNP discovery will be deposited in public database upon acceptance. The full  
320 SNP list can be found in the Figshare repository (accession number 10.6084/m9.figshare.7581581).

321

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456 **Table 1.** Summary of results from SNP discovery and quality control filtering for SNP selection in 326  
457 whole-genome sequenced individuals from three farmed Nile tilapia (*Oreochromis niloticus*) populations.

Chr <sup>a</sup>	Initial Number of SNP	First common filter <sup>b</sup>	Population specific filters <sup>c</sup>			Last common Filter <sup>d</sup>
			POP A	POP B	POP C	
<b>Mito<sup>e</sup></b>	1671	1596	0	0	0	0
<b>LG1</b>	779881	708381	6768	31144	18046	2045
<b>LG2</b>	891811	797928	7765	32143	12664	1881
<b>LG3a</b>	626179	539885	6583	17175	9991	714
<b>LG3b</b>	3491391	2943480	22758	47028	29881	1823
<b>LG4</b>	1099494	983024	11401	38514	23132	2150
<b>LG5</b>	792857	706930	6497	26718	11529	1811
<b>LG6</b>	1247549	1112177	10300	47240	23576	2560
<b>LG7</b>	1326771	1191205	16277	61818	24935	3549
<b>LG8</b>	834050	752753	10587	35798	17751	1803
<b>LG9</b>	842634	757126	7171	27404	13135	1518
<b>LG10</b>	767756	689181	8389	29290	13115	1774
<b>LG11</b>	904270	813207	8839	31506	16411	2074
<b>LG12</b>	1178720	1051496	12065	39106	18552	2181
<b>LG13</b>	797314	722353	7152	29682	18623	1819
<b>LG14</b>	960008	861238	11750	39640	17386	2176
<b>LG15</b>	991645	895658	10517	32442	19090	2002
<b>LG16</b>	1319450	1181919	12268	49954	25188	2447
<b>LG17</b>	1023476	919757	9226	40200	25573	2502
<b>LG18</b>	1111651	988338	10630	39619	18910	2082
<b>LG19</b>	624640	563099	7781	26643	15163	1799
<b>LG20</b>	762408	685589	7586	27546	13409	2013
<b>LG22</b>	1202188	1062702	13440	42665	18451	2016
<b>LG23</b>	1229543	1099828	11769	43157	25489	2610

<b>US<sup>f</sup></b>	5149044	4386247	24031	50640	31645	2651
<b>Total</b>	<b>29956401</b>	<b>26415097</b>	<b>261550</b>	<b>887072</b>	<b>461645</b>	<b>50000</b>

458 <sup>a</sup> Chromosome

459 <sup>b</sup> First common quality control filtering using all populations including SNP *excluded by genotypes*

460 *quality* < 15 and minimum site *quality* < 40

461 <sup>c</sup> Population specific quality control filtering including removing SNP with missing genotypes >0.60,

462 minor allele frequency (MAF) <0.01, Hardy-Weinberg Equilibrium (HWE) P-value < 1e-06, Illumina

463 score <0.8 and at least one Mendelian error in POP B.

464 <sup>d</sup> Last common quality control filtering retaining markers with unique position in the genome,

465 prioritizing SNP with MAF > 0.05 in the three commercial populations and evenly distributed across

466 the genome

467 <sup>e</sup> Mitochondria

468 <sup>f</sup> Unplaced scaffolds

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481 **Table 2.** Descriptive results of population genetic estimates and statistics for the different populations of  
482 farmed Nile tilapia using the 50 K SNP validation panel.

Population	HWE <sup>a</sup>		MAF > 0.05 <sup>b</sup>		MAF > 0.01 <sup>c</sup>		H <sub>O</sub> <sup>d</sup>	H <sub>E</sub> <sup>e</sup>
	n	%	n	%	n	%		
POP A	46,139	99.07	37,843	75.69	43,450	86.9	0.2011	0.2843
POP B	45,757	98.25	44,696	89.39	46,570	93.14	0.2497	0.3130
POP C	43,869	94.20	45,171	90.34	46,570	93.14	0.2463	0.3243

483 <sup>a</sup>SNPs in Hardy-Weinberg Equilibrium

484 <sup>b</sup>SNPs with Minor Allele Frequency > 0.05

485 <sup>c</sup>SNPs with Minor Allele Frequency > 0.01

486 <sup>d</sup>Observed heterozygosity

487 <sup>e</sup>Expected heterozygosity

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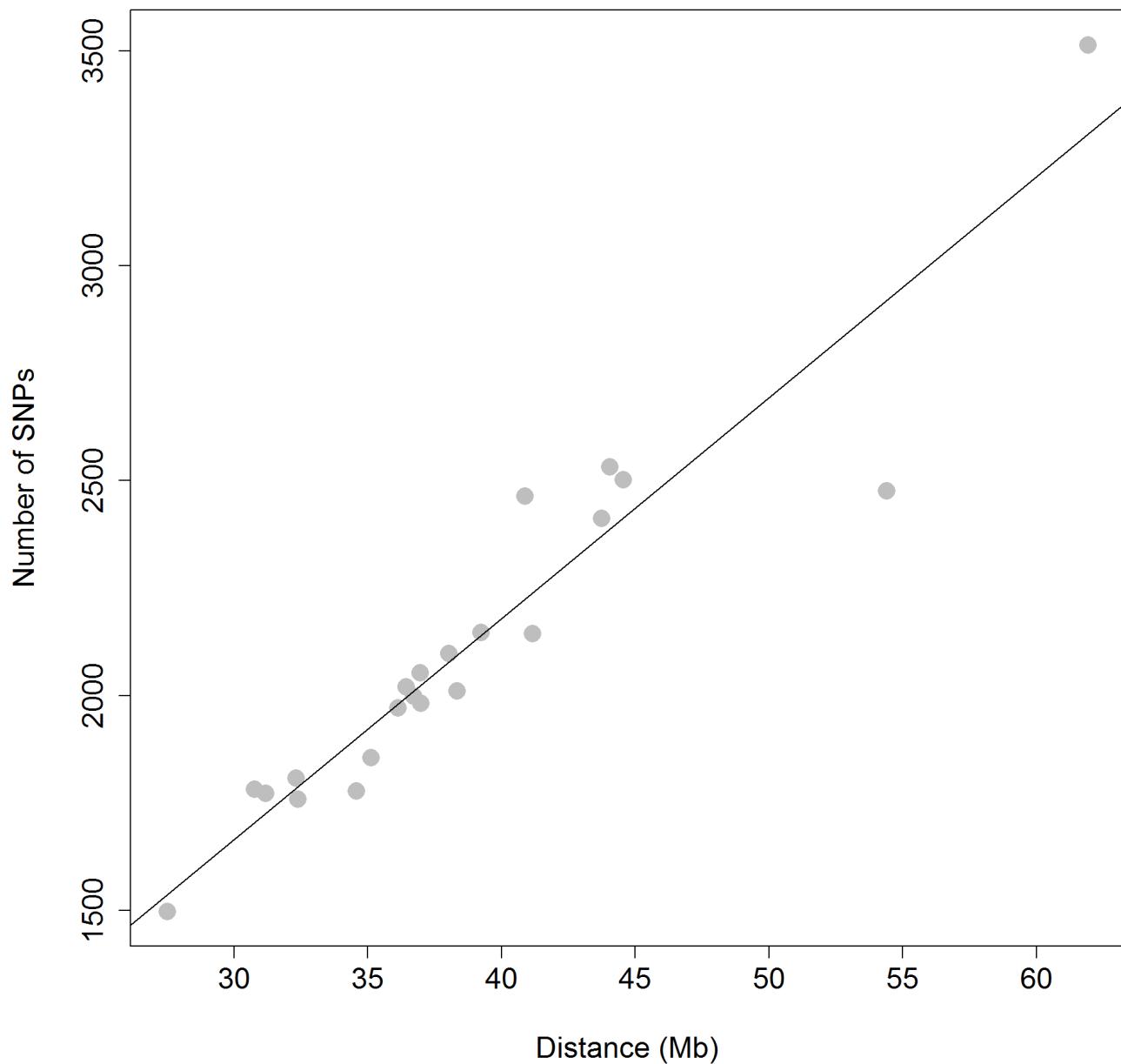
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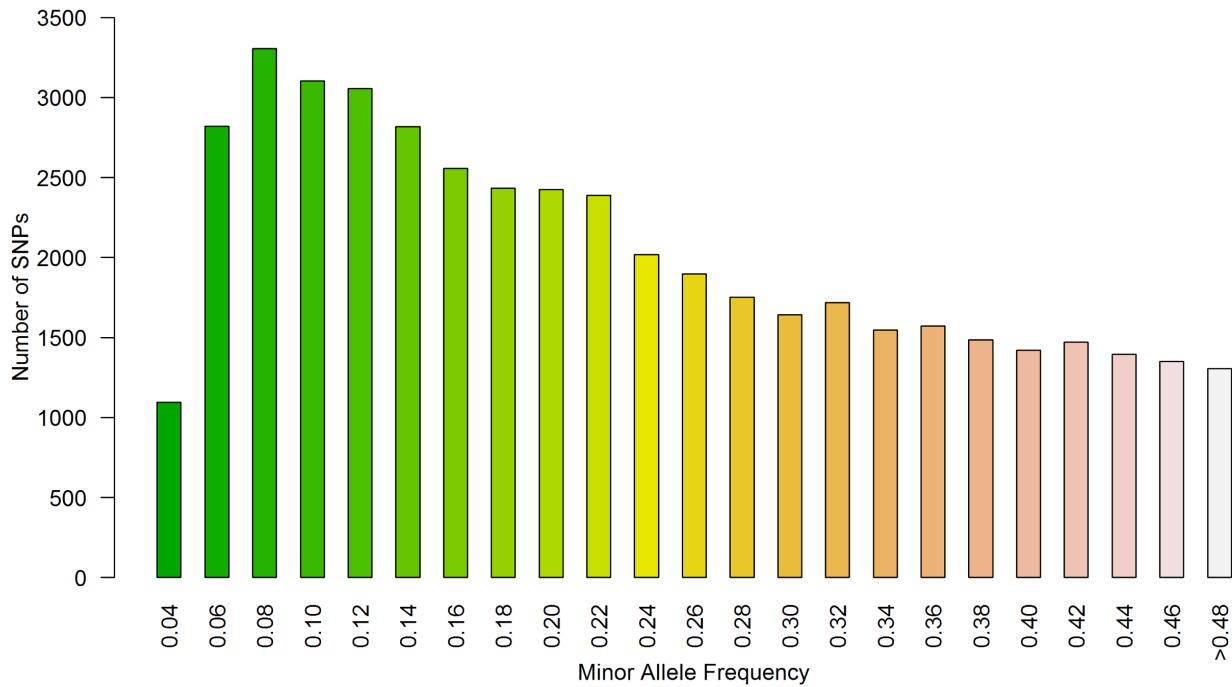
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503 **Fig. 1 Relationship between the number of SNPs and chromosome length.** Scatter plot of the number  
504 of SNPs per chromosome and the total chromosome length in Mb according to the assembly  
505 GCF\_001858045.1. The correlation coefficient between the number of SNPs and chromosome size is  $r =$   
506 0.95.

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509 **Fig. 2 Distribution of minor allele frequencies (MAFs).** Distribution of MAFs for the 50K SNP  
510 validation panel from 326 samples.

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