RESOURCE ARTICLE



Spatial dynamics and mixing of bluefin tuna in the Atlantic Ocean and Mediterranean Sea revealed using next-generation sequencing

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Gregory N. Puncher<sup>1,2,3*</sup> | Alessia Cariani<sup>1*</sup> | Gregory E. Maes<sup>4,5,6</sup> |
Jeroen Van Houdt<sup>5</sup> | Koen Herten<sup>5,6</sup> | Rita Cannas<sup>7</sup> | Naiara Rodriguez-Ezpeleta<sup>8</sup> |
Aitor Albaina<sup>9,10</sup> | Andone Estonba<sup>9</sup> | Molly Lutcavage<sup>11</sup> | Alex Hanke<sup>12</sup> |
Jay Rooker<sup>13,14</sup> | James S. Franks<sup>15</sup> | Joseph M. Quattro<sup>16</sup> | Gualtiero Basilone<sup>17</sup> |
Igaratza Fraile<sup>8</sup> | Urtzi Laconcha<sup>8,9</sup> | Nicolas Goñi<sup>8</sup> | Ai Kimoto<sup>18</sup> |
David Macías<sup>19</sup> | Francisco Alemany<sup>19</sup> | Simeon Deguara<sup>20</sup> | Salem W. Zgozi<sup>21</sup> |
Fulvio Garibaldi^{22} | Isik K. Oray^{23} | Firdes Saadet Karakulak^{23} | Noureddine Abid^{24} |
Miguel N. Santos<sup>25</sup> | Piero Addis<sup>7</sup> | Haritz Arrizabalaga<sup>8</sup> | Fausto Tinti<sup>1</sup>
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*These authors contributed equally.

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¹Department of Biological, Geological and Environmental Sciences/Laboratory of Genetics and Genomics of Marine Resources and Environment (GenoDREAM), University of Bologna, Ravenna, Italy

²Department of Biology, Marine Biology Research Group, Ghent University, Ghent, Belgium

³Department of Biology, University of New Brunswick, Saint John, NB, Canada

⁴Centre for Sustainable Tropical Fisheries and Aquaculture, Comparative Genomics Centre, College of Science and Engineering, James Cook University, Townsville, Qld, Australia

⁵Centre for Human Genetics, Genomics Core, KU Leuven - UZ Leuven, Leuven, Belgium

⁶Laboratory of Biodiversity and Evolutionary Genomics, University of Leuven (KU Leuven), Leuven, Belgium

⁷Department of Life & Environmental Sciences (DISVA), University of Cagliari, Cagliari, Italy

⁸Marine Research Division, AZTI Tecnalia, Pasaia, Gipuzkoa, Spain

⁹Laboratory of Genetics Faculty of Science & Technology, Department of Genetics, Physical Anthropology and Animal Physiology, University of the Basque Country (UPV/EHU), Leioa, Spain

¹⁰Environmental Studies Centre (CEA), Vitoria-Gasteiz City Council, Vitoria-Gasteiz, Spain

¹¹School for the Environment and Large Pelagics Research Center, University of Massachusetts, Boston, Gloucester, MA, USA

¹²Fisheries and Oceans Canada, St. Andrews Biological Station, St. Andrews, NB, Canada

¹³Department of Marine Biology, Texas A&M University at Galveston, Galveston, TX, USA

¹⁴Department of Wildlife and Fisheries Sciences, Texas A&M University, College Station, TX, USA

¹⁵Gulf Coast Research Laboratory, Center for Fisheries Research and Development, University of Southern Mississippi, Ocean Springs, MS, USA

¹⁶Department of Biological Sciences, University of South Carolina, Columbia, SC, USA

¹⁷National Research Council, Institute for Marine and Coastal Environment, Detached Unit of Capo Granitola, Trapani, Italy

 $^{^{\}rm 18}{\rm National}$ Research Institute of Far Seas Fisheries, Shizuoka, Japan

¹⁹Instituto Español de Oceanografía, Centro Oceanográfico de Baleares, Palma, Spain

 $^{^{\}rm 20}{\rm Federation}$ of Maltese Aquaculture Producers (FMAP), Valletta, Malta

²¹Marine Biology Research Center, Tripoli-Tajura, Libya

²²Department of Earth, Environmental and Life Sciences, University of Genoa, Genova, Italy

²³Faculty of Fisheries, Istanbul University, Laleli-Istanbul, Turkey

²⁴National Institute of Fisheries Research, Regional Centre of Tangier, Tanger, Morocco

²⁵Instituto Português do Mar e da Atmosfera, Olhão, Portugal

Correspondence

Gregory N. Puncher, University of New Brunswick, Saint John, NB, Canada. Email: gregory.puncher@unb.ca

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Abstract

The Atlantic bluefin tuna is a highly migratory species emblematic of the challenges associated with shared fisheries management. In an effort to resolve the species' stock dynamics, a genomewide search for spatially informative single nucleotide polymorphisms (SNPs) was undertaken, by way of sequencing reduced representation libraries. An allele frequency approach to SNP discovery was used, combining the data of 555 larvae and young-of-the-year (LYOY) into pools representing major geographical areas and mapping against a newly assembled genomic reference. From a set of 184,895 candidate loci, 384 were selected for validation using 167 LYOY. A highly discriminatory genotyping panel of 95 SNPs was ultimately developed by selecting loci with the most pronounced differences between western Atlantic and Mediterranean Sea LYOY. The panel was evaluated by genotyping a different set of LYOY (n = 326), and from these, 77.8% and 82.1% were correctly assigned to western Atlantic and Mediterranean Sea origins, respectively. The panel revealed temporally persistent differentiation among LYOY from the western Atlantic and Mediterranean Sea ($F_{ST} = 0.008$, p = .034). The composition of six mixed feeding aggregations in the Atlantic Ocean and Mediterranean Sea was characterized using genotypes from medium (n = 184) and large (n = 48) adults, applying population assignment and mixture analyses. The results provide evidence of persistent population structuring across broad geographic areas and extensive mixing in the Atlantic Ocean, particularly in the mid-Atlantic Bight and Gulf of St. Lawrence. The genomic reference and genotyping tools presented here constitute novel resources useful for future research and conservation efforts.

KEYWORDS

mixed-stock analysis, origin assignment, population structure, reduced representation sequencing, single nucleotide polymorphisms, *Thunnus thynnus*

1 | INTRODUCTION

Characterization of wide and small-scale population connectivity patterns of large fish species with extensive migrations remains one of the key challenges preceding the definition of reliable conservation and management units. The International Commission for the Conservation of Atlantic Tunas (ICCAT) manages the Atlantic bluefin tuna (BFT; Thunnus thynnus) as two stocks, each composed of a single population with the following spatial dynamics: a western group that spawns in the Gulf of Mexico and forages in the northwestern Atlantic Ocean and an eastern group that spawns in the Mediterranean Sea and forages in the Mediterranean and northeastern Atlantic Ocean (Fromentin & Powers, 2005). Several tagging and otolith studies have demonstrated that these two stocks mix extensively and form large feeding aggregations in the Atlantic Ocean, particularly in the vicinity of the Mid-Atlantic Bight (Block, Dewar, & Blackwell, 2001; Block, Teo, & Walli, 2005; Wilson et al., 2011). This evidence, combined with results from various otolith chemistry and

genetic studies, suggests that the population structure of BFT is much more complex than the current two-stock management model (Aranda, Abascal, Varela, & Medina, 2013; Cermeño, Quílez-Badia, & Ospina-Alvarez, 2015; Fraile, Arrizabalaga, & Rooker, 2014; Galuardi et al., 2010; Riccioni et al., 2010, 2013; Rooker, Secor, De Metrio, Kaufman, et al., 2008; Rooker, Secor, De Metrio, Schloesser, et al., 2008; Rooker et al., 2014). The resolution of the species' population structure and mixed-stock analysis is a high priority for managers and an overdue improvement for the suboptimal assessment models currently employed (ICCAT 2011). It is estimated that the eastern stock is much more productive (10×) than the western stock, and as such, the risk of overexploitation of western BFT in mixed fisheries is much higher (ICCAT 2015).

Genetic tools used for population structure studies are able to infer the spatial dynamics, migratory movements and kinship of commercially targeted species in order to delineate management units (Ovenden, Berry, Welch, Buckworth, & Dichmont, 2015). These details are essential for accurate estimations of individual stock

contributions to mixed fisheries, which are particularly vulnerable to mismanagement due to a tendency for overexploitation of less productive stocks (Habicht, Seeb, Myers, Farley, & Seeb, 2010; Taylor, McAllister, Lawson, Carruthers, & Block, 2011; Ying, Chen, Lin, & Gao, 2011). In the past 15 years, several molecular techniques have been employed in an effort to gain a better understanding of the population structure and spatial dynamics of BFT (see ICCAT 2013 for a summary).

Sequences from the mitochondrial DNA control region (mtDNA CR) provided the first evidence of differentiation between the western Atlantic and Mediterranean stocks (Alvarado-Bremer, Naseri, & Ely, 1999); however, the effectiveness of this marker was called into question by a later study that used samples collected during multiple seasons (Ely et al., 2002). Several studies published since then have provided a variety of estimates of differentiation of the two stocks using allozymes (Pujolar, Roldán, & Pla, 2003), mtDNA CR sequences and microsatellites (Boustany, Reeb, & Block, 2008; Carlsson, McDowell, Carlsson, & Graves, 2007), while other studies have not found any significant differentiation (Alvarado Bremer, Viñas, Mejuto, Ely, & Pla, 2005). Compared with the estimated mean and median F_{ST} values of 57 marine fish species (0.062 and 0.020, respectively; Waples, 1998), the aforementioned estimates of differentiation for the eastern and western stocks of BFT have been much lower (fixation indices ranging from 0.005 to 0.012).

Several studies have provided evidence of varying levels of differentiation between groups of BFT within the Mediterranean Sea across a range of geographic scales using mtDNA CR sequences and microsatellites (Carlsson et al., 2004, 2007; Riccioni et al., 2010, 2013). Conversely, several other studies have been unable to detect any genetic structure in the Mediterranean Sea using mtDNA CR sequences and microsatellites (Boustany et al., 2008; Vella et al., 2016; Viñas et al., 2011), EST-derived microsatellites (Riccioni et al., 2017) and single nucleotide polymorphisms (SNPs; Antoniou, Kasapidis, Kotoulas, Mylonas, & Magoulas, 2017).

In addition to the molecular markers selected, several important factors need to be considered when investigating the genetic structure of long-lived migratory fishes with ocean-wide distributions. Comparisons between BFT captured in adult feeding aggregations are likely to generate inaccurate and inconsistent allele frequencies, as they are likely to contain a varying proportion of individuals from the eastern and western stocks. As such, investigations into the genetic population structure of BFT should be limited to larvae and/or small young-of-the-year (YOY) captured in spawning areas and include samples from multiple years to avoid false positives for divergence due to sampling bias, temporal variation in the survival of larvae and sweepstake effects (Larson & Julian, 1999; Maes, Pujolar, Hellemans, & Volckaert, 2006).

The goal of this study was to develop a panel of SNPs capable of distinguishing BFT larvae and YOY (LYOY) collected in spawning and nursery areas in the western Atlantic Ocean and Mediterranean Sea using reduced representation sequencing (RRS) of DNA libraries. SNPs are commonly used for population genomic analyses and molecular ecology studies due to their high abundance in any given

genome, transferability between laboratories, high throughput potential and low genotyping error (Elshire et al., 2011: Hauser, Baird, Hilborn, Seeb, & Seeb, 2011; Hess, Matala, & Narum, 2011; Kelley, Brown, Overgaard Therkildsen, & Foote, 2016; Seeb et al., 2011). They have been used to investigate marine fish population structure (Albaina et al., 2013; Diopere et al., 2017; Jackson et al., 2014; Milano et al., 2014; Wang et al., 2016) and seafood traceability (Nielsen et al., 2012; Stokstad, 2010), providing the scientific premises for the monitoring of migratory stocks and further development of DNA-based forensics in fisheries law enforcement. The developed panel of SNPs was then used to differentiate medium and large size adults of eastern and western origin captured at several feeding aggregations in the Atlantic Ocean and Mediterranean Sea. This study is the most geographically comprehensive genetic analysis of BFT to date, with samples coming from as far away as Libya, Turkey, the Gulf of Mexico and the Gulf of St. Lawrence.

2 | MATERIALS AND METHODS

2.1 | SNP discovery and validation

2.1.1 | Tissue sampling and DNA extractions

A total of 105 larvae and 450 YOY of BFT were captured at four spawning areas in the Mediterranean Sea (Balearic Sea, Tyrrhenian Sea, Strait of Sicily and Levantine Sea) and western Atlantic Ocean (Gulf of Mexico spawning site and adjacent waters off Cape Hatteras) between 2007 and 2012 during the late spring to early autumn months (Figure 1, Table 1). During 2009 and 2012, samples were provided by multiple suppliers in the Gulf of Mexico (2009), Tyrrhenian Sea (2012) and Levantine Sea (2012). Larvae were collected using plankton nets and stored in 96% ethanol, according to standard larval survey practices. In addition to morphological identification, all larvae were identified to species level using molecular techniques described in Puncher et al. (2015). YOY was captured with hand and trolling lines. The size range (straight fork length) of YOY samples (Min. = 12 cm, Max. = 55 cm, the Avg. = 31.3 \pm 6.5 cm; Table S1) corresponds with an estimated age of 2 weeks-12 months (Avg. = 3 months; Diaz & Turner, 2007; Restrepo et al., 2010). All research was coordinated through the Atlantic-wide Research Programme for Bluefin Tuna (GBYP) for which ICCAT issued a recommendation allowing the parties involved in this research to collect and sacrifice young BFT for the purposes of genetic research as well as ship samples from one country to another (Cert. No. ICCAT RMA12-049; ICCAT Circulars 2296-12, 2180-14, 3203-15).

Approximately 20 mg of muscle tissue or fin-clip from each of the YOY and complete or partial (either caudal fin or eyeball) larvae was digested overnight in a proteinase K solution. Genomic DNA was extracted from all tissue and purified using Promega's Wizard SV96 Genomic DNA Purification kit and vacuum manifold following the manufacturer's protocols. DNA was eluted with 60 μl of distilled water preheated to 60°C.

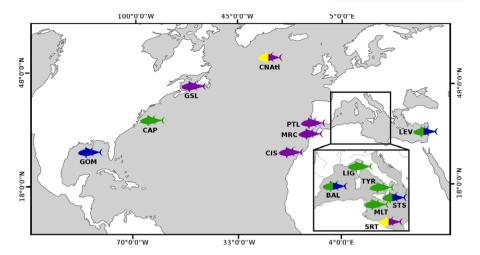


FIGURE 1 Sampling locations of Atlantic bluefin tuna of various size classes used in this study. Inset: detailed map of sampling locations in the western and central Mediterranean Sea. Larvae = Blue; Young-of-the-year = Green; Yellow = Medium adult; Purple = Large adult. Sample sizes, year of sampling and code definitions are detailed in Table1

2.1.2 | Restriction site-associated DNA sequencing

Extracted DNA was first digested with the ApeKI restriction enzyme (RE), which has a 5-bp recognition site (5′GCWGC - 3′CGWCG), followed by ligation of standard Illumina sequencing adaptors with RE sequencing overhangs (Elshire et al., 2011). Libraries were sequenced at the Genomics Core Facility (KULeuven) using a HiSeq2500 (Illumina) 100-bp paired-end module, multiplexing 188 individuals on each sequencing lane with barcodes developed within the project (Table S2). To achieve a sufficient depth of sequencing for reliable genotype calling, 316 individuals were resequenced (274 of which had fewer than 500K reads). After resequencing, 90% of all individuals had >500K reads and the average total read count for all individuals was 1,673,390 \pm 1,407,838. The average read count of resequenced individuals increased from 253,139 to 2,006,368 (Table S2).

2.1.3 | SNP discovery

Barcode and low-quality edges were removed from the raw Illumina DNA sequence data, and de novo SNP calling was initially performed using the Universal Network Enabled Analysis Kit pipeline (UNEAK, Lu et al., 2013). Among the 324,433 loci analysed, SNPs that were not recognized among more than 70% of all individuals within each sample were removed. Allele frequencies among loci and individuals and genotype coverage were then analysed using the Adegenet package in R (Jombart & Ahmed, 2011) and GENETIX (Belkhir, Borsa, Chikhi, Raufaste, & Bonhomme, 2004), and an excess of heterozygosity was observed for the majority of the remaining loci. In case this phenomenon was a consequence of low individual read coverage, we performed additional allele frequency analysis using individuals with more than 2 million read counts each. The de novo SNP discovery pipeline was then abandoned for reasons outlined in the supplementary materials (Text S1).

In response to the aforementioned challenge, a genomic reference was produced by assembling data generated by genome sequencing of a single individual captured offshore from the Balearic Islands (NCBI BioProject PRJNA432036). The genomic library was sequenced in 2 × 75-bp paired-end mode on a HiSeq 2000 (Illumina). A total of 94,000,207 paired-end reads were produced and were quality trimmed to 71 bp. ABYSS (version 1.3.7; Simpson et al., 2009) was used to create a de novo assembly with standard parameters and an optimal k-mer size of 31. This resulted in an assembly of 6,309,067 contigs and a genome size of 944 Mb. As quality control for the assembled genomic contigs, the original data were mapped using BOWTIE2 (version 2.2.0; Langmead & Salzberg, 2012) in the local modus. Summary statistics from the de novo assembly of the genomic data are provided in Tables S3 and S4.

Barcodes and adaptors were then trimmed from the raw pairedend sequence data of all 555 individuals (allowing for one sequencing error in the barcodes and enzyme restriction site) and sequences de-multiplexed using GBSX (Herten, Hestand, Vermeesch, & Van Houdt, 2015). Overlapping read pairs of each individual in the data set were then merged using FLASH (version 1.2.7; Magoč & Salzberg, 2011) with default parameters, producing longer sequences (180 bp), which resulted in much improved mapping quality and data confidence (Table S5, overall retention of 88.9% of reads). The data for all 555 individuals were then pooled according to the main geographic sampling areas, four from the Mediterranean (Balearic Sea, Tyrrhenian Sea, Strait of Sicily and Levantine Sea) and one from the western Atlantic Ocean (merging Gulf of Mexico and Cape Hatteras; Table 1). The five sequence data pools were mapped against the genomic reference using the end-to-end modus in BOWTIE2 (version 2.2.0; Langmead & Salzberg, 2012).

SNP calling was performed simultaneously for the five pools in one run using FREEBAYES (version 9.9.10; Garrison & Marth, 2012), based solely on merged single mapping and similar settings to the human genome project (mapping quality >20 and base quality >15;

TABLE 1 Details of Atlantic bluefin tuna used in this study for: (i) in silico variant discovery, (ii) validation of a 384 SNP genotyping panel and then selection of the 96 SNPs panel and (iii) large-scale genotyping with 96 SNP panels. Integers in parentheses in the final column represent individuals removed from analyses due to low genotyping success (<70% SNPs genotyped)

					Sample sizes			
Sampling location	Area code	Age class	Year	Sample code	SNP discovery	SNP validation and selection	96 SNP typing	
Western Atlantic Ocean								
Gulf of Mexico	GOM (WAtl)	Larvae (V)	2007	WAGM07V	11	2		
		Larvae (V)	2008	WAGM08V	14	8		
		Larvae (V)	2009	WAGM09bV	25	10		
		Larvae (V)	2009	WAGM09V	24	8		
		Larvae (V)	2010	WAGM10V			40 (4)	
Cape Hatteras	CAP (WAtl)	YOY (0)	2008	WACH080	16	16		
Gulf of St. Lawrence	GSL	Large Adult (L)	2013	WAGSL13L			23 (1)	
Central North Atlantic Oc	cean							
Central North Atlantic	CNAtl	Medium Adult (M)	2012	CACA12M			24	
		Large Adult (L)	2012	CACA12L			47 (1)	
Eastern Atlantic Ocean								
Canary Islands	CIS	Large Adult (L)	2013	EACI13L			24	
Morocco	MRC	Large Adult (L)	2012	EAMO12L			24	
		Large Adult (L)	2013	EAMO13L			19	
Portugal	PTL	Large Adult (L)	2011	EAPO11L			23	
Mediterranean Sea								
Balearic Sea	BAL	YOY (0)	2009	WMBA090	41			
		YOY (0)	2010	WMBA100	45			
		YOY (0)	2011	WMBA110	40	24		
		YOY (0)	2012	WMBA120	40			
		Larvae (V)	2012	WMBA12V			40	
		YOY (0)	2013	WMBA130			34	
Ligurian Sea	LIG	YOY (0)	2013	WMLI130			33	
Tyrrhenian Sea	TYR	YOY (0)	2011	WMTY110	37			
		YOY (0)	2012	WMTY120	40	24		
		YOY (0)	2012	WMTY12b0	40			
		YOY (0)	2013	WMTY130			40	
Malta	MLT	YOY (0)	2013	CMMA130			40	
Strait of Sicily	STS	Larvae (V)	2008	CMSI08V			25 (2)	
		YOY (0)	2011	CMSI110	21	21		
		Larvae (V)	2012	CMSI12V	21			
		YOY (0)	2012	CMSI120	40	27		
		YOY (0)	2013	CMSI130			40	
Gulf of Sirte	SRT	Medium Adult (M)	2012	CMSY12M			24	
		Large Adult (L)	2012	CMSY12L			24	
Levantine Sea	LEV	Larvae (V)	2011	EMLS11V	10			
		YOY (0)	2011	EMLS110	29	27		
		YOY (0)	2012	EMLS120	40			
		YOY (0)	2012	EMLS12b0	21			
		YOY (0)	2013	EMLS130			40	
				TOTAL	555	167	564 (8)	

http://www.genome.gov/10000923). Reads that were mapped more than once and/or discordantly to the genomic reference were discarded. Results were analysed using pairwise comparisons of allele read counts per locus. Several studies have demonstrated that this PoolSeq approach is a reliable method for SNP discovery, which offers faithful estimates of population allele frequencies and low false-positive rates (Futschik & Schlötterer, 2010; Ozerov, Vasemägi, Wennevik, Niemelä, et al., 2013; Zhu, Bergland, González, & Petrov, 2012). The genomic coordinates and corresponding coverage statistics for the targeted regions were retrieved using BEDTOOLS (version 2.17.0; Quinlan & Hall, 2010). Depth of coverage for each location in the genomic reference was normalized for each sample (all individuals captured at a sampling location during the same survey) using the following formula:

 $\label{eq:Normalized depth} Normalized depth location = Location depth \\ \times \frac{\mbox{Highest total read number across all samples}}{\mbox{Total number of reads within sample}}$

Locations from the genomic reference with a normalized depth of \geq 50 reads, among all samples, were selected for SNP identification. By performing SNP calling simultaneously on the mapped data of all samples, SNP information for all candidate positions in all samples was obtained, including positions that were homozygous for certain samples.

2.1.4 | SNP selection and validation

A minimum read count of 18 per locus and pooled geographic sample was used as a threshold for confidence in SNP identity. Variants identified in silico were first selected for validation based upon minimum coverage and Delta value selection as described in Ozerov, Vasemägi, Wennevik, Diaz-Fernandez et al. (2013; Delta = [pA_i pA_i], where pA_i and pA_i are the frequencies of allele A in the ith and jth populations). A panel of 384 loci was then identified based upon compliance with Biomark™ HD Fluidigm array design criteria (minimum flanking sequence length of 60 bp and <65% GC content). The panel was validated using a subset of the 555 individuals initially processed for SNP discovery, selecting 167 BFT representative of the five geographical sampling areas (western Atlantic Ocean, Balearic Sea, Tyrrhenian Sea, Strait of Sicily and Levantine Sea; Table 1). SNP typing was carried out with the Biomark™ HD platform (Fluidigm) and run with ROX as the passive reference dye and FAM and HEX SNPtype[™] assays for specific allele calling. Individuals with low genotyping coverage (<70% SNPs genotyped) and SNP loci with an excess of missing data (>10%) and with minor allele frequencies of < 0.01 among all samples were removed from the data set. To select SNPs optimal for discrimination of eastern and western spawning groups, as well as for within-Mediterranean discrimination, a reduced number of candidate SNPs was then identified with BELS (Bromaghin, 2008), which uses a backward elimination procedure that excludes each locus from the data set in order of increasing impact on differentiation. As a performance measure "maximize mean mixture-composition estimation accuracy" was selected, population

proportions were random, mixture sample size was 200, and each simulation used 250 replications. A final panel of 96 SNPs was selected and compared with sequences contained in the NCBI database using the blastn algorithm in BLAST (Altschul, Gish, Miller, Myers, & Lipman, 1990).

2.2 | Assignment to origin and mixed-stock analysis

2.2.1 | Collection, DNA extractions and SNP typing of additional LYOY and adult samples

The efficacy of the selected 96 SNP panel for stock assignment analysis was then assessed by genotyping a second group of LYOY from the western (n = 40) and eastern spawning areas (n = 292; Table 1). DNA was extracted from 232 adult BFT from two different size classes (medium adults: 25–100 kg; large adults: >100 kg; Table 1) and six geographic locations throughout the species' range (Figure 1) and genotyped using the methods outlined above. Combined with the samples used for SNP validation (n = 167), the total number of individuals genotyped using the 96 SNP panel was 731 (Figure 1; Table 1).

2.2.2 | Genetic diversity and outlier detection

Kinship analysis was conducted for each temporal and spatial sample of LYOY. Relatedness was evaluated using ML-RELATE (Kalinowski, Wagner, & Taper, 2006), which assigns each pair of individuals to a range of relationships (parent–offspring "PO," full-sibling "FS," half-sibling "HS" and unrelated "U"), the most likely of which was determined by comparing a putative relationship (the highest likelihood value) with an alternative relationship (the second highest likelihood value) using 1,000 simulations. Similar sibship inference was performed with COLONY v. 2.0.5.8 (Jones & Wang, 2010) using the full likelihood method for a medium length run with medium likelihood precision. Genotyping error rates were set at 0.01 per SNP site with no allelic dropout, and polygamy among noninbreeding males and females was assumed. To be classified as belonging to one of the three relatedness categories (unrelated, half-sibs and full-sibs), dyads had to have ≥50% probability of belonging to a given category.

After relatedness assessment, BFT captured in the Gulf of Mexico between 2007 and 2009 were pooled for all calculations, due to the limited number of larvae in each sample. Values of expected and observed heterozygosity and the inbreeding coefficient ($F_{\rm IS}$; Weir & Cockerham, 1984) were calculated for each sample using ARLEQUIN v3.5.2 (Excoffier & Lischer, 2010), allowing for 10% missing data. Departure from Hardy–Weinberg equilibrium (HWE) was assessed for each locus in each group as well as across the entire data set using the exact test implemented in GENEPOP v.4.2 (Raymond & Rousset, 1995; Rousset, 2008) using 1,000K Markov chain steps and 100K dememorization steps. Linkage disequilibrium (LD) tests were conducted with Genepop using default values (100 batches of 1,000 iterations each). The false discovery rate (B-H) approach (Benjamini & Hochberg, 1995) was used to analyse the resulting LD p-values

for false positives using the SGOF+ program (Carvajal-Rodriguez & de Uña-Alvarez. 2011).

Outlier loci were identified using the Beaumont and Nichols (1996) FDIST approach, implemented in Lositan (Antao, Lopes, Lopes, Beja-Pereira, & Luikart, 2008) with a 0.995 confidence interval, 50K simulations and a false discovery rate of 0.1. The hierarchical Bayesian method described in Beaumont and Balding (2004), implemented in Bayescan software (Foll & Gaggiotti, 2008), was also used to detect loci with strong support for selection ($-\log_{10}PO > 1$) using default settings (5K iterations, 50K burn-in length, 20 pilot runs of 5,000).

2.2.3 | Stock discrimination

Pairwise comparisons of F_{ST} values between all samples were executed using FSTAT v.2.9.4 (Goudet, 1995) and between western and eastern pooled LYOY groups with ARLEQUIN v.3.5.2, testing for significance with an adjusted nominal level of α = 0.05 and 10K permutations. The performance of the 96 SNP panel, to differentiate western and eastern LYOY pooled groups, was once again analysed with BELS, using the same settings described above. Genetic heterogeneity and differentiation of LYOY were estimated in ARLEQUIN, using several analyses of molecular variance (AMOVA) to examine multiple levels of variance in the data set. The AMOVAs were performed with 10K permutations and a significance level of 0.05. The following three groupings were explored by assigning samples to various a priori groups: (i) western Atlantic Ocean (Gulf of Mexico and Cape Hatteras) vs. Mediterranean Sea, (ii) temporal replicates pooled by geographic region (see Table 1) and (iii) western Mediterranean Sea (Balearic Sea) vs. central western Mediterranean (Tyrrhenian Sea and Ligurian Sea) vs. central Mediterranean (Strait of Sicily and Malta) vs. eastern Mediterranean Sea (Levantine Sea). Additional analyses were performed using genotypes from two subsets of SNPs selected for trans-Atlantic and intra-Mediterranean differentiation, respectively.

Genetic admixture and stock discrimination were further investigated using STRUCTURE v2.3.4 (Pritchard, Stephens, & Donnelly, 2000), which employs a Bayesian model-based individual clustering method. STRUCTURE provides an option to use geographic sampling locations as prior information to assist with the clustering of individuals in data sets that have genuine population structure, but the signal is too weak for standard structure models to detect (Hubisz, Falush, Stephens, & Pritchard, 2009). The LOCPRIOR model is suitable when one is assuming that the sampling locations may be informative about ancestry, as is the case with LYOY captured close to their origin. STRUCTURE analysis was conducted with the following settings: for each simulation of K (1-10), five independent replicates were run, with a 100K burn-in period and 500K Markov Chain Monte Carlo reps after burn-in, Admixture Model, LOCPRIOR using the geographical regions defined in Table 1 and lambda inferred for each run. A plot of the mean likelihood L(K) and variance per K value from each of the STRUCTURE analyses was obtained using Structure Harvester (Earl & von Holdt, 2012) and used to infer the optimal *K* value or number of ancestral clusters.

2.2.4 | Stock assignment and mixed-stock analysis

A hold-out group (sensu Anderson, 2010) of 332 LYOY not used for SNP selection or validation was genotyped (Section 2.2.1) and assigned to eastern or western origin using both a leave-one-out self-assignment approach as well as using a reference data set (training set sensu Anderson, 2010) composed of genotypes from the 167 LYOY used for SNP selection and validation (Table 1). The GENECLASS2 software (Piry et al., 2004) was employed using the Rannala and Mountain (1997) criterion (0.05 threshold). Individual results are reported as assignment scores, which are the likelihood of belonging to x population divided by the sum of likelihoods of belonging to each of the stocks considered. Individuals with an assignment score <70% were considered "poorly assigned" and are included in the results to demonstrate the proportional contribution of individuals falling within this criteria from each stock. A quality index was also calculated, reflecting the mean value of the scores of each individual in the population it belongs to. The probability of assignment to spawning area of origin was calculated with the Monte Carlo resampling probability computation using the Paetkau, Slade, Burden, and Estoup (2004) simulation algorithm and 10,000 simulated individuals in GENECLASS2, using the same query and LYOY reference data sets.

Medium and large BFT adults from six feeding aggregations in the Atlantic Ocean and Mediterranean Sea (N=230; 2 individuals were removed due to low genotyping success) were then assigned to western Atlantic and Mediterranean Sea spawning areas with GENECLASS2, using the complete LYOY data set (N=493) and the Rannala and Mountain (1997) criterion (0.05 threshold). Overall probability of assignment to either spawning area was calculated using the same settings described above and the complete LYOY data set.

The 100% simulations analysis tool in ONCOR was used to examine the potential accuracy of the mixture analysis using the sample sizes of the LYOY data set, 100 simulations and 200 fish sampled from a virtual fishery. Mixture analysis was then performed on each adult feeding aggregation using the conditional maximum-likelihood method (Millar, 1987) used by ONCOR (Kalinowski, Manlove, & Tape, 2007) to estimate mixture proportions. Both the LYOY and adult data sets were resampled with 20,000 bootstraps to calculate confidence interval values.

2.3 | Estimates of effective population size

The effective population size ($N_{\rm e}$) of the eastern and western pooled groups was calculated using NEESTIMATOR v. 2.0 (Do et al., 2014). A series of five $N_{\rm e}$ calculations were conducted using equal sample sizes by randomly selecting 80 individuals from the eastern pooled group using the *sample* function in R. All tests were run using the linkage disequilibrium method (Waples & Do, 2010), random mating model and without any allele frequency restrictions.

3 | RESULTS

3.1 | SNP discovery and validation

Statistics related to the assembly of the genomic contigs used as a reference for mapping and variant calling are featured in Tables S3 and S4. After implementation of the mapping and normalized depth filtering criteria, SNP calling resulted in 2,467,103 SNP positions, of which 236,332 intersected with the >50× coverage mapped locations. After filtering for only bi-allelic variants (among the five geographical pools), a total of 184,895 SNPs were found to be shared among all five geographic sampling areas (minimum 50× coverage across all pools with fragments present with at least 10× coverage in each pool). When selecting variants for SNP assays, preference was given to loci that had Delta values between 1.0 < 0.7 (high; n = 64), 0.7 < 0.6 (mid; n = 99) and 0.6 < 0.5 (low; n = 221) between the five geographical pools.

Technical validation of the 384 variants identified in silico (SNP flanking sequences available at https://doi.org/10.5061/dryad.2f7v9) had a 75% conversion rate to functional Fluidigm genotyping assays. Further assessment of the remaining 288 loci using genotype data from 167 individuals revealed 77 candidate SNPs with low polymorphism (MAF < 0.01) across the entire data set or a high frequency of missing data (>10%). Although these values were likely to change with an increase in sample size, these 77 loci were removed in favour of SNPs that are consistently informative. This left a set of 211 SNPs for which the average percentage of missing data per SNP locus was 2.57% (genotype matrix available at https://doi.org/ 10.5061/dryad.2f7v9). According to BELS results, a total of 59 SNPs were selected for discrimination of LYOY from the western Atlantic Ocean and the Mediterranean Sea and another 48 SNPs for discrimination of sampling locations within the Mediterranean (11 of which had already been selected for trans-Atlantic discrimination). In total, 27 loci of the selected 96 were found to share >80% BLAST similarity scores with previously published gene sequences. Annotation details for the 96 SNPs selected for the final genotyping panel are provided in Table S6.

3.2 Assignment to origin and mixed-stock analysis

3.2.1 | Genetic diversity and outlier detection

The selected 96 SNP panel provided high genotyping success rate among most of the 731 individuals (99%) genotyped. Only eight individuals were removed from the data set due to low genotyping coverage (Table 1; genotype matrix available at https://doi.org/10.5061/dryad.2f7v9). A single locus was removed from analysis, due to low polymorphism throughout the entire data set, resulting in a final set of 95 loci (58 candidate markers for trans-Atlantic discrimination and 48 for intra-Mediterranean differentiation).

Kinship analysis of LYOY suggested a very low proportion of related pairs (above the half-sibling level) among samples (Table S7). In fact, the majority of dyads were identified as UNRELATED (range

91.4%–100%), with only a few dyads classified as full- and or half-siblings.

Observed and expected heterozygosity among LYOY ranged from 0.291 to 0.343 and from 0.316 to 0.346 per sample, respectively (Table S8). Observed F_{IS} values were low and not significant, agreeing with results from the aforementioned kinship analysis. None of the 95 SNPs significantly deviated from the HWE. After applying the false discovery rate to all 3,160 tests, there was no evidence of linkage disequilibrium among loci. Moreover, each of the selected SNP loci was contained in separate contigs, and as such, physical linkage can also be excluded. When comparing eastern and western LYOY pooled groups, outlier analysis (using Lositan) revealed that three loci (SNP_161, SNP_158, SNP_010) are candidates for positive selection ($F_{ST} = 0.122$, 0.082 and 0.076, respectively). A BLAST of the sequences flanking each of these SNPs provided a single match (SNP_158) with anaphase promoting complex subunit 2 sequences of several other teleosts (90% nucleotides matched in a ~145-bp sequence; Table S6).

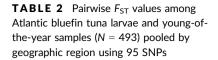
3.3 | Genetic differentiation among LYOY samples

Based on pairwise comparisons using the 95 SNP panel, there was no differentiation among samples collected from different years at the same sampling location (Table S9). For this reason, multiyear geographic samples were combined. BFT caught during different sampling seasons in Cape Hatteras (YOY) and the Gulf of Mexico (larvae), including larvae collected in the Gulf of Mexico in 2010 (WAGM10V), which were not used for SNP panel selection, showed no genetic differentiation, providing further support to pool temporal and geographical samples in this region. Within the Mediterranean Sea, the only pairwise comparison that resulted in significant differentiation was between Malta 2013 and the Levantine Sea 2011 $(F_{ST} = 0.010; Table S9)$. After pooling temporal replicates by geographic area, western Atlantic Ocean samples significantly differed from all Mediterranean Sea samples ($F_{ST} = 0.007-0.012$, $p \le .002$; Table 2). BELS analysis revealed that only 24 loci were required for optimal differentiation of these two groups of samples (Table S6). Analyses of pairwise comparisons using 58 and 24 SNP panels showed consistent differentiation between western and eastern BFT $(F_{ST} = 0.012-0.024, p < .002; F_{ST} = 0.032-0.040, p < .002, respec$ tively). Pairwise comparisons using the 95 and 48 SNP panels detected no temporally stable significant differences between Mediterranean Sea samples (data not shown).

Among the different pooled groups analysed using AMOVAs, the only scenario that provided significant statistical support was that of separate Mediterranean Sea and Gulf of Mexico populations (Tables S10 and S11). Pooled groups of LYOY from these two locations were significantly different using panels of 95, 58 and 24 loci ($F_{ST} = 0.008$, p = .034; $F_{ST} = 0.014$, p = .035 and $F_{ST} = 0.034$, p < .001, respectively). Differentiation of eastern and western pooled groups remained significant when assessing only LYOY that were not used for the SNP panel development and validation (N = 326, after removal of six individuals with low genotyping

	WAtl	BAL	LIG	TYR	MLT	STS	LEV
WAtl		≤0.002*	≤0.002*	≤0.002*	≤0.002*	≤0.002*	≤0.002*
BAL	0.010*		0.155	0.414	0.005	0.136	0.871
LIG	0.008*	0.001		0.512	0.064	0.279	0.433
TYR	0.007*	0.000	0.001		0.207	0.098	0.202
MLT	0.012*	0.004	0.004	0.001		0.024	0.076
STS	0.009*	0.001	0.001	0.001	0.004		0.455
LEV	0.010*	0.000	0.002	0.001	0.002	0.000	

 $F_{\rm ST}$ values are below and associated *p*-values above the diagonal. Significant pairwise comparisons are marked with an asterisk. Sample sizes, age class, year of sampling and area code definitions are detailed in Table 1.



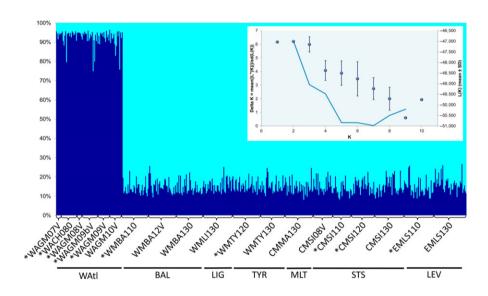


FIGURE 2 STRUCTURE plot of 493 young Atlantic bluefin tuna samples (larvae and young-of-the-year) genotyped at 95 SNPs. Individual columns represent the proportion of membership from two inferred ancestral clusters (K = 2). Inset: Plot of the mean likelihood L(K) and variance per K value from all STRUCTURE analyses. Asterisks indicate samples used for the 96 SNPs panel selection. Sample sizes, age class, year of sampling and code definitions are detailed in Table 1

coverage), using these same panels ($F_{ST} = 0.004$, p = .009; $F_{ST} = 0.008$, p < .001, $F_{ST} = 0.024$, p < .001, respectively).

Bayesian cluster analysis suggested that modern BFT stocks are descendants of two distinct ancestral populations (K = 2), as currently represented by the samples collected in the western Atlantic Ocean/Gulf of Mexico and Mediterranean Sea (Figure 2).

3.4 | Population assignment and admixture analysis

3.4.1 | Larvae and young-of-the-year

Among the LYOY used for SNP selection, 75.0% from the western Atlantic (n = 33/44) and 77.2% from the Mediterranean (n = 95/123) were correctly assigned to origin (Table S12). A total of 11.3% and 14.6% of individuals captured in the western Atlantic and Mediterranean, respectively, were considered poorly assigned (assignment scores <70%; Table S12).

Assignment of LYOY not included in the SNP discovery, validation and selection processes (hold-out group sensu Anderson, 2010; N=326) demonstrated consistent efficacy of the panel. However, this hold-out data set did yield a lower quality index of 70.9%, with 74.8% of individuals correctly assigned to origin. Probability

computation with Monte Carlo resampling assigned 83.4% of individuals from this hold-out group to correct origin with an average probability of assignment of 82.9%. Overall 66.7% and 71.7% of individuals captured in western and eastern spawning areas were correctly assigned to origin with average assignment scores of 89.6% and 87.1%, respectively (Table S12). A total of 19.4% and 20.3% of individuals assigned to the western and eastern stocks were considered poorly assigned (Table S12). Similar results were obtained when assigning individuals in the hold-out group to their origin using the reference data set (n = 167; training set sensu Anderson, 2010; Table S13). A total of 72.7% of individuals in the hold-out group was assigned to origin, while 14.4% was considered poorly assigned.

An analysis of the SNP panel's ability to assign LYOY to origin using all LYOY individuals (N=493) was also conducted; however, in an effort to avoid high grading bias, summary statistics reported in Table 3 are limited to individuals not used for SNP selection (N=326; refer to Table S14 for assignment scores of all LYOY samples). Among these individuals, 66.7% of western and 72.4% eastern LYOY were correctly assigned to origin with average assignment scores of 86.4% and 89.0%, respectively (Table 3). The proportion of poorly assigned individuals collected

TABLE 3 Assignment of Atlantic bluefin tuna larvae and young-of-the-year (N = 326) not included in the SNP panel selection, captured in spawning and nursery areas in the western Atlantic (WAtl) and Mediterranean Sea (Med) using genotypes from 95 SNPs

Sample Location	Sample code	N	WAtI A.S. >70%	WAtl A.S. <70%	Avg. A.S. to WAtl	Med A.S. >70%	Med A.S. <70%	Avg. A.S. to Med
Western Atlantic	WAGM10V	36	66.7	11.1	86.4	11.1	11.1	72.7
Balearic Sea	WMBA12V	40	7.5	10.0	69.9	72.5	10.0	90.9
Balearic Sea	WMBA130	34	8.8	8.8	68.2	73.5	8.8	92.4
Ligurian Sea	WMLI130	33	12.1	3.0	73.7	69.7	15.2	86.2
Tyrrhenian Sea	WMTY130	40	5.0	12.5	67.0	75.0	7.5	89.4
Malta	CMMA130	40	12.5	2.5	73.6	70.0	15.0	88.2
Strait of Sicily	CMSI08V	23	4.3	13.0	66.2	78.3	4.3	91.6
Strait of Sicily	CMSI130	40	17.5	7.5	76.6	67.5	7.5	88.9
Levantine Sea	EMLS130	40	12.5	5.0	78.3	75.0	7.5	85.4
Total WAtl		36	66.7	11.1	86.4	11.1	11.1	72.7
Total Med		290	10.3	7. 6	72.2	72.4	9.7	89.0

Individuals with an assignment score (A.S.) <70% were considered "poorly assigned," and data are reported separately. All values, aside from sample size (N), are reported as percentages. All calculations were made using GENECLASS2. See Table 1 for sampling locations and sample codes.

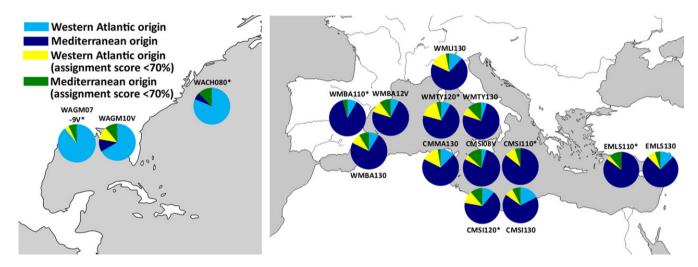


FIGURE 3 Atlantic bluefin tuna larvae and young-of-the-year (*N* = 493) captured in spawning and nursery areas in the western Atlantic and Mediterranean Sea assigned to natal origin using genotypes from 95 SNPs. Individual assignment was determined using likelihood estimates calculated in GeneClass2. Individuals with an assignment score of <70% were considered "poorly assigned." Asterisks indicate samples used for the 96 SNPs panel selection. Sample sizes, age class, year of sampling and code definitions are detailed in Table 1

from the western Atlantic Ocean (22.2%) was higher than those collected in the Mediterranean Sea (17.3%; Figure 3; Table 3).

3.4.2 | Adults

The average probability of assignment of all adult individuals (N=230) to spawning areas of origin was estimated at 86.8%. Average assignment scores for adults to the western Atlantic Ocean and Mediterranean Sea were 75.2% and 86.6%, respectively, with an overall average of 84.5% (Table 4; Figure S3). The proportion of individuals assigned to either western or eastern spawning areas was calculated and plotted on a map illustrating geographic distribution of mixing rates (Figure 4; Table 4).

An evaluation of the power of assignment revealed that the actual sample sizes from the western Atlantic Ocean (n=80) and Mediterranean Sea (n=413) included in the LYOY data set and used for genetic stock identification of adults were likely to provide $87.9 \pm 6.4\%$ and $98.7 \pm 1.1\%$ accuracy (Table S15). Mixture analysis of feeding aggregations provided similar results (Table S16) as the individual assignment of adults; however, the 95% confidence intervals provided by the ONCOR analysis are extremely wide and should be interpreted with caution, as these measures are sensitive to unbalanced sample sizes. For example, the proportion of individuals assigned to western origin is much reduced, in comparison with other methods, at locations where the western fraction is low. This is likely due to the lower number of individuals from the western Atlantic in the LYOY data set.

TABLE 4 Assignment of adult Atlantic bluefin tuna (N = 230), captured in various feeding aggregations, to natal origin in the western Atlantic (WAtl) and Mediterranean Sea (Med) using genotypes from 95 SNPs

Sampling Location	Sample Code	N	WAtI A.S. >70%	WAtl A.S. <70%	Avg. A.S. to WAtl	Med A.S. >70%	Med A.S. <70%	Avg. A.S. to Med
Gulf of St. Lawrence	WAGSL13L	22	36.0	14.0	78.9	41.0	9.0	87.6
Central North Atlantic	CACA12L	46	13.0	4.0	84.4	59.0	24.0	81.0
Central North Atlantic	CACA12M	24	8.0	4.0	70.2	67.0	21.0	86.2
Canary Islands	EACI13L	24	8.0	13.0	71.9	75.0	4.0	90.9
Morocco	EAMO12L	24	8.0	8.0	70.4	79.0	4.0	88.5
Morocco	EAMO13L	19	16.0	16.0	71.4	63.0	5.0	91.1
Portugal	EAPO11L	23	0.0	4.0	51.3	70.0	26.0	84.4
Gulf of Sirte	CMSY12L	24	4.0	0.0	70.0	83.0	13.0	89.0
Gulf of Sirte	CMSY12M	24	4.0	4.0	69.2	79.0	13.0	87.5
Total		230	10.9	7.0	75.2	67.8	14.3	86.6

Individuals with an assignment score (A.S.) <70% were considered "poorly assigned," and data are reported separately. All values, aside from sample size (N), are reported as percentages. All calculations were made using GENECLASS2. See Table 1 for sampling locations and sample codes.

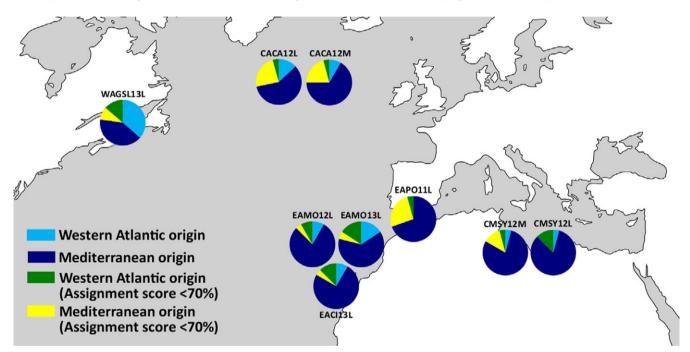


FIGURE 4 Adult Atlantic bluefin tuna (N = 230) captured in various feeding aggregations assigned to natal origin in the western Atlantic and Mediterranean Sea using genotypes from 95 SNPs. Individual assignment was determined by comparing genotypes of adults to a reference data set of 493 larvae and young-of-the-year using likelihood estimates calculated in GeneClass2. Individuals with an assignment score of <70% were considered "poorly assigned." Sample sizes, age class, year of sampling and code definitions are detailed in Table 1

Both methods used to analyse the composition of adult feeding aggregations revealed that the central Atlantic Ocean, Gulf of St. Lawrence and coastal Morocco are areas hosting the highest amount of mixing between stocks (Table 4). Conversely, the samples from Portugal are entirely of Mediterranean Sea origin. The contribution of adult BFT from each stock to the feeding aggregation in Morocco appears to vary slightly from 1 year to the next.

3.5 | Estimates of effective population size

The effective population size estimated for the pooled western Atlantic group (3437.2, confidence interval = $469.6 - \infty$) was lower

than the $N_{\rm e}$ estimated for the pooled eastern Atlantic group (4682.2, confidence interval = 1840.7 $-\infty$). Results from tests using equal sample sizes (n = 80) were so variable (Table S17) that no conclusions can be drawn from their consideration.

4 | DISCUSSION

4.1 De novo SNP discovery

The difficulties that were encountered during our initial SNP discovery efforts using a de novo approach were not unique. Several studies have shown that *ApeKI* library preparations, coupled with

Illumina sequencing, results in a low, biased and error-prone coverage of genome sequences (De Donato, Peters, Mitchell, Hussain, & Imumorin, 2013; Lu et al., 2013). The restriction site targeted by ApeKI (GC[A/T]GC) is relatively short and contains a degenerate base, allowing frequent fragmentation of the genome. Short fragments can easily be incorrectly aligned, resulting in paralogs that appear as heterozygous loci. Moreover, this can result in a high number of missing data in post-sequencing data sets, irrespective of the quality of template DNA (De Donato et al., 2013). Using these methods researchers can expect to have ~30% missing data for any given individual in a data set, even with high depth coverage (Beissinger et al., 2013). For this reason, many recent GBS studies have used alternative restriction enzymes with longer base pair cut sites (Pstl, EcoT22I) for studies of wild nonmodel species (Alcaide, Scordato, Price, & Irwin, 2014; De Kort et al., 2014; Li et al., 2014). A more in-depth discussion about challenges associated with the de novo SNP discovery methods that were used is featured in the supplementary materials (Text S1).

4.2 | SNP genotyping

As we chose to pool all sequence data according to geographical origin of individuals, the focus on allele frequencies was shifted to overall read counts. This is not a new approach and has been advised for tasks requiring reduced representation library construction and less ambitious sequencing efforts (Gautier et al., 2013). Pooling of genotype data from individuals sequenced using NGS platforms can provide accurate estimates of population allele frequencies with similar accuracy to individual-based analyses; however, it also reduces capacity for detection of rare alleles (Cutler & Jensen, 2010).

4.3 | Population assignment and mixture analysis

Kinship analysis indicated that the relations among LYOY collected from each location were sufficiently distant so as not to cause concern for duplication of genotypes within the data set. The sampling techniques used (sampling on multiple dates at different locations) avoided the collection of patches of closely related individuals which can confuse population structure patterns and lead to under-estimation of genetic diversity.

Juvenile bluefin tuna of >12 months age (>53 cm) have been observed to cross the Atlantic Ocean (Cort & Liorzou, 1995; Fraile et al., 2014; Rooker, Secor, De Metrio, Kaufman, et al., 2008). In this study, we have used younger YOY as genetic references to avoid the inclusion of migrants in our spawning location reference data set (Table S1). For example, the age of the YOY captured in Cape Hatteras was approximately 2–5 months (Min. = 12 cm; Max. = 37 cm; Avg. = 29.6 cm; Diaz & Turner, 2007). It is unlikely that these young fish have crossed the Atlantic Ocean, as large fast-swimming adult BFT require 3–4 months to complete the journey (Block et al., 2001; Mather, 1980; Wilson et al., 2011).

The range of F_{ST} values that we have calculated (0.009–0.034) based on the separation of the eastern and western LYOY samples

are similar to those calculated by a previous study that used mtDNA CR sequences (Φ_{ST} = 0.013) and microsatellites (F_{ST} = 0.006; Carlsson et al., 2007). However, it must be noted that the SNPs used to calculate F_{ST} values were selected based on their capacity to differentiate the two stocks. The first effort to characterize BFT populations in the western and eastern Atlantic Ocean using SNPs produced the most significant evidence for differentiation of populations to date (Albaina et al., 2013). In that study, 17 SNPs revealed significant differentiation of YOY captured in the western Atlantic Ocean and western Mediterranean Sea ($F_{ST} = 0.116 \pm 0.078$, p < .01) as well as juveniles from the Bay of Biscay (F_{ST} = 0.120 \pm 0.091, p < .01). These F_{ST} estimates are an order of magnitude higher than previous estimates using other molecular approaches. Due to the limited number of western Atlantic individuals used in their study (N = 15), the results should be verified with an increased sample size. Moreover, at least 80 SNPs are recommended for the detection of low levels of differentiation among highly mobile species (Hauser et al., 2011; Ryman et al., 2006).

Variability of $F_{\rm ST}$ values among locations sampled over multiple years, throughout the entire data set, is indicative of temporal genetic variability among cohorts. This suggests a reconsideration of the conclusions reached by previous studies that have used single-season sample collections and supports the argument that multiyear sampling is a necessity for studies investigating the population structure of BFT as well as other migratory marine fishes.

Considering the highly migratory nature of juvenile and adult BFT, the weak signal of differentiation (low yet statistically significant F_{ST} values) between the two stocks is not surprising. Only a handful of individuals (<1%) migrating between spawning areas each generation can eliminate most genetic evidence of stock structure (Waples, 1998). The F_{ST} values generated here are relatively low when compared to other taxa; however, these values are consistent with what is to be expected from a highly migratory marine species with a larval phase and recently generated populations (Alvarado Bremer et al., 2005; Ward, Woodwark, & Skibinski, 1994).

The broad distribution of individual assignment scores that we calculated can be expected from highly variable candidate source populations (see Tables S10 and S11 for within-population variation; Ogden & Linacre, 2015). To improve the assignment of juvenile and adult BFT to Gulf of Mexico and Mediterranean Sea spawning areas, additional LYOY are required from the Gulf of Mexico in order to account for this variation.

4.4 Assignment power and confidence

The extent to which individual LYOY were successfully assigned to origin (79.3% for all LYOY, 77.3% for the hold-out group, considering all assignment scores) is similar to the results of studies using various properties of otoliths for assignment. Rooker, Secor, Zdanowicz, De Metrio, and Relini (2003) were able to assign YOY to eastern and western origins with an 85% success rate, using trace elements (Li, Mg, Ca, Mn, Sr and Ba) in otoliths. Using stable isotopes (δ^{13} C and δ^{18} O), Schloesser, Neilson, Secor, and Rooker (2010) and Rooker

et al. (2014) were able to assign YOY from eastern and western nurseries to their origin with overall success rates of 84% and 83%. With the same markers, Hanke, Busawon, Rooker, and Secor (2016) also encountered an overall assignment success rate of 83% when assigning adult BFTs caught in the Northwest Atlantic Ocean from 1975 to 1977 and 2011 to 2013. Similarly, Brophy et al. (2015) were able to distinguish between eastern and western BFT using otolith shape, once again with an 83% success rate. The results of these five studies combined suggest the existence of an unknown source of error that accounts for a 16%–20% misclassification rate.

While genetic and otolith analyses provide similar assignment success rates, it is important to note that only genetic analysis is effective for the identification of butchered market tuna with heads removed. Moreover, the collection of tissues suitable for genetic analysis (scales, fin clips, spines, muscle) is much easier, noninvasive and does not require the killing of fish. Genetic assignment results are informative at an evolutionary scale, while other methods (e.g. morphology) provide information that is mainly informative at an ecological timescale and on each individual's pre/postreproductive movements. Multi data source information should be ultimately combined or integrated rather than compared.

In comparison with population assignment analysis, results from the admixture analysis appear to underestimate the proportion of individuals from western spawning/nursery areas. Given the much higher productivity of the eastern stock and the tendency of maximum-likelihood estimations to eclipse estimates of less productive stocks by more dominant ones (Fraile et al., 2014; Millar, 1990), we discuss the results from both population assignment and admixture analyses.

4.5 | Evidence of mixing

Both tagging and otolith chemistry indicate that stocks mix extensively in the Atlantic Ocean, particularly around the mid-Atlantic Bight, the central north Atlantic and off the coasts of Morocco and western Europe (Block et al., 2001, 2005; Fraile et al., 2014; Galuardi et al., 2010; Rooker, Secor, De Metrio, Kaufman, et al., 2008; Rooker et al., 2014; Wilson et al., 2011). The extent of these oceanic movements appears to be dependent upon BFT origin and ontogeny (Block et al., 2005; Mather, 1980). Although adults are capable of crossing the Atlantic in 3-4 months (Block et al., 2001; Wilson et al., 2011, 2015), significant numbers of juvenile BFT (>3 up to 44 kg) also make the trip (Graves et al., 2015; Mather, 1980; Rooker, Secor, De Metrio, Kaufman, et al., 2008; Rooker et al., 2014). According to various tagging studies, BFT from the east are more likely to cross the Atlantic than their counterparts in the west (Block et al., 2005; Taylor et al., 2011). Our results are corroborated by previously published synopses stating that stock mixing is highest in the western Atlantic Ocean, with the eastern stock subsidizing the western stock by 10%-50% (Taylor et al., 2011).

Evidence of BFT reproduction taking place in a third, previously undocumented spawning area in the Slope Sea (western

Atlantic Ocean) has been provided by multiple sources (Baglin, 1976: Goldstein, Heppell, Cooper, Brault, & Lutcavage, 2007: Heinisch, Rosenfeld, Knapp, Gordin, & Lutcavage, 2014; Mather, Mason, & Jones, 1974, 1995); however, the extent to which this occurs remains unknown. Recently. Richardson et al. (2016) provided strong evidence of this third spawning area by collecting larvae in the western Slope Sea. Combined with the presence of reproductively active medium-sized BFT (120-220 cm) in the same area, this finding supports a scenario of size-structured spawning migrations in the western Atlantic spawning groups, as suggested by life history modelling (Chapman, Jørgensen, & Lutcavage, 2011). In such a scenario, medium-sized BFT may feed and spawn in the Slope Sea, while large adults from the eastern and western stocks have a wider spatial distribution that includes feeding areas in the North Atlantic Ocean and spawning areas in the Mediterranean Sea and Gulf of Mexico, respectively. Moreover, BFT captured in the Mediterranean Sea and identified as western in origin may come from the Slope Sea and not just from the Gulf of Mexico. This might account for the relatively weak levels of differentiation between the currently recognized western and eastern stocks. These revelations may provide an explanation for the aforementioned 16%-20% of individuals that cannot be assigned to either spawning group using otolith characteristics, or the techniques described herein. Moreover, lower assignment scores in some individuals may be an indication that gene flow is occurring in the Slope Sea.

4.5.1 Northwestern Atlantic Ocean

The size fraction of YOY BFT collected from Cape Hatteras (fork length = 29.6 \pm 5.7 cm. total weight = 0.5 \pm 0.2 kg) was likely too small to contain any of the juvenile trans-Atlantic migrants documented by various studies (Mather, 1980; Rooker, Secor, De Metrio, Kaufman, et al., 2008; Rooker, Secor, De Metrio, Schloesser, et al., 2008). Based on our results, the adult fishery in the Gulf of St. Lawrence appears to contain a significant fraction of eastern migrants (>50%). A recent study, using otolith microchemistry assignment techniques, uncovered similarly high numbers of eastern migrants in the Gulf of St. Lawrence (26%), trans-Atlantic Bight (35%) and off the coast of Newfoundland (41%) and Nova Scotia (22%; Hanke et al., 2016 and Table S18). Although proportions may vary with time and sampling sites, estimates of the annual contribution of eastern BFT to the catch show a large increase in each of these regions in 2014 (Table S18). Rooker et al. (2014) have also estimated the proportion of eastern migrants in the Gulf of St. Lawrence mixed stock to be as high as $44.0 \pm 16.8\%$. Conversely, an earlier analysis of archived BFT otoliths caught in the same area suggested that eastern migrants were all but absent from the Gulf of St. Lawrence during 1975-1984 and 2006-2007 (Schloesser et al., 2010).

Catches of BFT in the Gulf of St. Lawrence indicate a steady increase in abundance since 2010 (Vanderlaan, Hanke, Chassé, & Neilson, 2014) with a growing proportion of the catch composed of

smaller fish. A tendency for fish size to be inversely related to the probability of eastern origin has been demonstrated (Hanke et al., 2016). This putative growing proportion of eastern migrants in the Gulf of St. Lawrence could be an indication of a poleward shift (Dell, Wilcox, Matear, Chamberlain, & Hobday, 2015; MacKenzie, Payne, Boje, Høyer, & Siegstad, 2014; Robinson, Hobday, Possingham, & Richardson, 2015) of eastern migrants in the western Atlantic Ocean.

4.5.2 Central North Atlantic

Genotyping of samples collected in the central North Atlantic Ocean revealed that the majority of large (83.0% overall; 59.0% when individuals with assignment scores of <70% are excluded) and medium-sized (88.0% overall; 67.0% when individuals with assignment scores of <70% are excluded) adults were of eastern origin. Using fish caught in the same area and season, Brophy et al. (2015) estimated that the mixed fishery contained approximately 94 \pm 7% BFT of eastern origin. An analysis of otoliths from 202 adults captured in the same area in 2010–2011 concluded that eastern migrants constitute between 79.5 \pm 6.5% (based on maximum-likelihood estimations) and 90.7 \pm 5.3% (2011 alone) of the mixed stock in that region (Rooker et al., 2014).

4.5.3 Northeastern Atlantic and Mediterranean Sea

Similar to some otolith-based studies (Rooker, Secor, De Metrio, Kaufman, et al., 2008; Rooker et al., 2014), we have also identified a small number of western-type individuals among samples collected in the Mediterranean Sea. These individuals could be incorrectly assigned, due to the low number of individuals contained in the western Atlantic LYOY reference pool (N=80; Tables 3 and S13–15). Alternatively, they could be an indication of previously undocumented mixing within spawning areas, as suggested by the two western-type adults caught in the Gulf of Sirte (Libya). The highest number of western migrants among the adult individuals collected in the eastern Atlantic was captured off the coast of the Canary Islands (21%) and Morocco (16%–32%). Previously, it has been estimated that $6.1\pm4.7\%$ to $22.6\pm7.0\%$ of BFT captured off the coast of Morocco are of western origin (Brophy et al., 2015; Rooker et al., 2014).

4.6 | Interannual variation

The results from the population assignment and mixture analysis of adult samples suggest that the contribution of each stock to feeding aggregations varies over time and across age classes. A previous genetic study determined that the mixed fishery in the North Atlantic Ocean, south of Iceland, contains BFT of significant genetic and morphological heterogeneity (Carlsson, McDowell, Carlsson, Ólafsdóttir, & Graves, 2006). Distinct groups of BFT appear to enter North Atlantic waters at different times throughout the fishing

season. Our results may provide an origin for these different cohorts of tuna. Rooker et al. (2014) observed considerable interannual variability in the proportion of western BFT in the central North Atlantic Ocean, ranging from 36.1 \pm 9.6% in 2010 to 9.3 \pm 5.3% in 2011. Inability to take these spatiotemporal patterns of stock variation into consideration for policy design and management imperils the welfare of the species as well as the sustainable exploitation of both stocks (Taylor et al., 2011). Other fisheries that target Baltic Sea cod (Eero, Hemmer-Hansen, & Hüssy, 2014; Hüssy et al., 2013) and various salmon stocks on the west coast of North America (Hess, Whiteaker, Fryer, & Narum, 2014; Larson, Seeb, Pascal, Templin, & Seeb, 2014) already consider this information when assigning temporal and regional quotas. This would require that the routine monitoring of BFT fisheries by way of catch estimates and morphological measurements should also include aspects of genetic and otolith analyses.

4.7 | Effective population size

Our estimates of N_e for the eastern pooled group (4,682, CI = 1,841 – ∞) were slightly higher than other previously published short-term N_e estimates for the Adriatic Sea and Tyrrhenian Sea which were calculated with pseudo-likelihood (752, 95% CI = 429–1,853; 326, 95% CI = 193–695, respectively) and likelihood-based (682, 95% CI: 363–1575; 405, 95% CI = 221–974, respectively) methods using the mutation rate of eight microsatellites (Riccioni et al., 2010). Our estimates are similar to values calculated for another closely related tuna species occupying the same ecological niche in the North Atlantic Ocean, the albacore tuna, which were generated using a panel of 75 SNPs (5,466–23,330; Laconcha et al., 2015).

Caution should be taken when interpreting our estimations of $N_{\rm e}$, however, given the apparent admixture of the two groups and the assumption of closed populations when estimating $N_{\rm e}$ using the linkage disequilibrium method (Waples & England, 2011). Moreover, the inconsistency of the $N_{\rm e}$ estimates and large confidence intervals generated by subsets of 80 individuals from the eastern pooled group suggests that a higher sample size is required to generate accurate estimates.

5 | CONCLUSION

Developing genomic resources for Atlantic Bluefin tuna is of crucial importance for the continued modernization of techniques used for management of the species as well as other tuna species that are equally threatened by overexploitation. The genomic reference presented here is of use for the mapping and positioning of current and future genomic markers. Annotation and comparison of our reference against other tuna or teleost genomes will allow better functional characterization of tuna genes linked to conservation and relevant life history traits. Genomewide SNP data, and especially outlier loci, have broad applications in future studies, allowing for local analyses and the validation of population structure at

microscale levels. Moreover, the development of a database containing SNP allele frequencies will allow for additional membership analyses of LYOY captured at spawning locations and adults caught in mixed feeding aggregations. Finally, the various resources that we have developed can be used for seafood traceability and fisheries forensics applications (Bernatchez et al., 2017).

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DATA ACCESSIBILITY

All sequencing data have been published at the National Center for Biotechnology Information (NCBI) as BioProject PRJNA432036, including the newly assembled genomic reference as well as the raw reads from the reduced representation library sequencing 555 individual Atlantic Bluefin tuna (BioSamples SAMN08408093 - SAMN08408647). SNP flanking regions and alleles, together with individual genotypes, are available on the Dryad Digital Repository (https://doi.org/10.5061/dryad.2f7v9).

AUTHOR CONTRIBUTIONS

G.N.P., A.C., G.E.M., R.C., A.A., A.E., H.A. and F.T. performed the project development and design; G.N.P., A.C., G.E.M., J.V.H., K.H., R.C., A.A., A.E., M.L., A.H., J.R., A.K., N.R.-E., H.A. and F.T. performed the manuscript development; G.N.P., A.C., G.E.M., J.V.H. and K.H. performed the laboratory analysis; G.N.P., A.C., G.E.M., J.V.H., K.H., R.C., N.R.-E. and F.T. performed the data analysis; G.N.P., A.C., I.F., U.L. and N.G. performed the data management; A.A., A.E., M.L., A.H., J.R., J.S.F., J.M.Q., G.B., A.C., I.F., U.L., N.G., A.K., D.M., F.A., S.D., S.W.Z., F.G., I.K.O., F.S.K., N.A., M.N.S., P.A., H.A. and F.T. performed the sample collection.

ORCID

Gregory N. Puncher http://orcid.org/0000-0002-4372-8435

Alessia Cariani http://orcid.org/0000-0002-5513-0086

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