The genomic basis and environmental correlates of local adaptation in the Atlantic horse

mackerel (*Trachurus trachurus*)

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Abstract

Understanding how populations adapt to local environments is increasingly important to prevent

biodiversity loss due to climate change. Here we examined whole-genome variation of 12

Atlantic horse mackerel pool samples from the North Sea to north Africa, and the western

Mediterranean Sea. This marine migratory benthopelagic fish is one of the most widely

distributed and commercially important species in the eastern Atlantic. We found low population

structure at neutral loci, but high differentiation at adaptive loci distinguishing the western

Mediterranean and the North Seas from other Atlantic locations. Candidate genes distinctive of

the Mediterranean include a green-sensitive-opsin harbouring two missense mutations that might

fine-tune the spectral sensitivity to blue-green light conditions. Candidate genes characteristic of

the North Sea could play a critical role in cold tolerance (energy metabolism and cell membrane

structure) and increased sensitivity to odours, presumably to compensate reduced visibility in
turbid waters. We also discovered a massive chromosomal inversion (9.9 Mb) that underlies a
climate-related latitudinal cline with a break near mid Portugal. Genome-environment
association analysis indicated that sea water dissolved oxygen concentration and temperature are
likely the main environmental drivers of local adaptation. Our genomic data broadly supports the
current ICES stock divisions, but recommends revision of the western and southern stock
boundaries. We developed a reduced SNP panel that genetically discriminate the North Sea from
neighbouring populations. Our study highlights the importance of life history and chromosomal
inversions in adaptation with gene flow, and the complexity of evolutionary and ecological
processes involved in local adaptation.
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Keywords
Whole-genome resequencing, genomics, adaptation, management, conservation, marine fish.

1. Introduction
To which extent marine species with no evident barriers restricting gene flow show genetic
differentiation and local adaptation across geographic areas is a question of considerable interest
in evolutionary biology, conservation, and management (Palumbi, 1994). Resolving this question
has been challenging because several marine species exhibit large population sizes, high gene
flow, and minute genetic drift, resulting in low genetic differentiation that has been difficult to
resolve with neutral genetic markers (Hauser & Carvalho, 2008a).
Owing to advances in high-throughput sequencing, recent genomic studies screening
thousands to millions of genetic markers across the genome have revealed population structure
and selection signatures in species previously assumed to be panmictic (e.g. Atlantic herring
(Han et al., 2020)) or lowly structured (e.g. Atlantic cod (Barth et al., 2017), Atlantic halibut
(Kess et al., 2021)). Population structure in marine fish has been characterized by shifts in allele
frequencies at many small effect loci or fewer large effect loci (Gagnaire & Gaggiotti, 2016) and
in chromosomal rearrangements (Han et al., 2020). Moreover, genomic divergence has been
linked to ecological variation, for example, in migratory behavior (Kirubakaran et al., 2016),
seasonal reproduction (Lamichhaney et al., 2017; Petrou et al., 2021), or local adaptation along
environmental gradients (Fuentes-Pardo et al., 2019; Stanley et al., 2018). Therefore, a thorough

examination of genomic variation, including neutral and adaptive loci, can help identify distinct

biological units and genetic variants associated with adaptation to local environments,

knowledge of great interest in conservation and management, especially in the face of climate

change.

Fish stock identification is an important prerequisite for fisheries management (Cadrin &

Secor, 2009), however many exploited stocks have traditionally been defined according to

geographical and political features or regions rather than on a biological basis. As more

information becomes available, it is evident that the temporal and spatial distributions of most

fisheries resources are not aligned to these artificial divisions (Kerr et al., 2017) and that

biological populations are more dynamic and complex (Reiss, Hoarau, Dickey-Collas, & Wolff,

2009; Stephenson, 2002). Therefore, it is critical to identify the underlying population structure

of fisheries resources and use this information to identify the appropriate level at which to define

assessment and management units. It is also important to be able to assign individuals in mixed

survey and commercial catches to the population or assessment unit to which they belong

(Casey, Jardim, & Martinsohn, 2016; Hintzen et al., 2015).

The Atlantic horse mackerel (*Trachurus trachurus* Linnaeus, 1758) is a marine

benthopelagic shoaling species widely distributed in the east Atlantic, from Norway to southern

Africa (FAO Major Fishing areas 27, 34 and 47), and in the Mediterranean Sea (FAO Major

Fishing area 37) (Froese & Pauly, 2021). It inhabits a variety of environments and different

populations may be under contrasting selection pressures, making this species an ideal species

for the study of local adaptation. Horse mackerel are generally found in continental shelf waters

(100-200 m depth), but have also been observed in deeper (~1000 m) or near shore waters. The

species undertakes annual migrations between spawning, feeding and over-wintering areas

(Abaunza et al., 2003). Horse mackerel are considered to be asynchronous batch spawners with

indeterminate fecundity and are not known to be faithful to their original spawning grounds

(Gordo et al., 2008; Ndjaula, Hansen, Krüger-Johnsen, & Kjesbu, 2009). Eggs and larvae are

pelagic and are often found from the surface to 100 m deep on the continental shelf or near the

coast (Alvarez & Chifflet, 2012; Beveren, Klein, Serrão, Gonçalves, & Borges, 2016).

In European waters in the northeast Atlantic, horse mackerel are assessed and managed as

three separate stocks: the Western, the North Sea, and the Southern (Figure S1), which were

largely defined based on the results of the multidisciplinary HOMSIR project (Abaunza et al.,

2008). Populations inhabiting coastal waters along Morocco and Mauritania, in northwest Africa,

are considered a separate group, denominated the "Saharo-Mauritanian stock". However, they

are less studied and monitored than those in the north. The discreteness of the stocks and the

location and levels of mixing between them are unknown, which leads to uncertainty in the input

data for stock assessments. Previous genetic studies have indicated low genetic differentiation

between populations, and have provided inconclusive results in regards to population sub-

structuring beyond the three main stock divisions identified using traditional methods (Brunel et

al., 2016; Cimmaruta, Bondanelli, Ruggi, & Nascetti, 2008; Comesaña, Martínez-Areal, &

Sanjuan, 2008; Farrell & Carlsson, 2018; Healey et al., 2020; Kasapidis & Magoulas, 2008;

Mariani, 2012; Sala Bozano et al., 2015).

Given the elusive nature of the population structure of the Atlantic horse mackerel and its

ecological and commercial importance in the east Atlantic, the aims of this study were four-fold:

(i) Estimate the extent of genetic differentiation between populations based on whole-genome

sequencing; (ii) Identify the genetic basis of local adaptation; (iii) Identify which ecological and

evolutionary processes may have shaped patterns of population subdivision; (iv) Design a cost-

effective and highly informative SNP panel that can be used for future population studies and

genetic stock identification.

2. Materials and Methods

2.1 Sampling and DNA isolation

Samples were collected opportunistically between 2015 and 2017 through existing fisheries

surveys and from both target and non-target fisheries at 12 locations across the east Atlantic and

the western Mediterranean Sea (Figure 1). We aimed to collect spawning fish to ensure that

samples could provide a valid baseline (Figure S1). However, due to the opportunistic nature of

sampling this was not always possible (Table S1). Maturity stages were recorded by sample

collectors using a number of different maturity keys. Therefore, these were standardised to the

six-point international horse mackerel maturity scale (Table S2; (ICES, 2015)). A 0.5 cm^3 piece

of tissue was excised from the dorsal musculature of each specimen and stored at 4°C in absolute

ethanol. Total genomic DNA was extracted from the majority of samples by Weatherbys

Scientific Ltd, Ireland from 30 mg of tissue using sbeadex™ magnetic bead-based extraction

chemistry on the LGC Oktopure™ platform. The remaining samples were extracted using a

CTAB or Chelex and proteinase-K based extraction protocol (Table S1). The Chelex protocol

produced single-stranded DNA, whereas the other methods, double-stranded. DNA quantity was

measured with a NanoDrop ND-1000 spectrophotometer.

2.2 Pool library preparation and sequencing

We performed whole-genome sequencing of pooled DNA (pool-seq) to assess the genomic

variation among samples. This method provides population-level allele frequencies by

sequencing to a high depth a single-barcoded library prepared from a mixture (pool) of DNA of

individuals from a population (Schlötterer, Tobler, Kofler, & Nolte, 2014). DNA pools were

prepared by mixing equal amounts of DNA of 30 to 96 individuals collected in close spatial and

temporal proximity. DNA pools were quantified using a Qubit Fluorometer (Thermo Fischer

Scientific Inc) aiming to have at least 1.5 µg of DNA in 25-50 µL and were submitted to the

SNP&SEQ Technology Platform in Uppsala, Sweden, for library preparation and high-

throughput sequencing. A PCR-free Illumina TruSeq library with an insert size of 350 base pairs

(bp) was prepared for most pools, except for the ones extracted with Chelex, for which a Splinted

Ligation Adapter Tagging (SPLAT) library was used instead, as it is aimed for single-stranded

DNA (Raine, Manlig, Wahlberg, Syvänen, & Nordlund, 2017). Paired-end short reads (2x150

bp) were generated using an Illumina NovaSeq sequencer and S4 flow cells.

2.3 Read mapping and variant calling

Low quality bases, sequencing adapters, and short reads were removed from the raw reads using

Trimmomatic v.0.36 (Bolger, Lohse, & Usadel, 2014). Clean reads were mapped against the

Trachurus trachurus genome assembly (Accession: GCA_905171665.1, (Genner & Collins,

2022)) using *bwa-mem* 0.7.17 (Li, 2013). A pilot examination of read mapping statistics

indicated that the temporal replicates from northern and southern Portugal (NPT2, SPT2) might

be affected by technical artefacts (Figure S2). Hence, these samples were excluded from

genome-wide analyses, but were included in analyses focused on specific loci (e.g., in the

examination of allele frequency patterns).

Variant calling was performed using the algorithm *UnifiedGenotyper* of *GATK* v3.8

(McKenna et al., 2010). Biallelic SNPs were retained and various filters were applied to remove

spurious markers (see Supplementary Data for details). The resulting high-quality SNPs were

used in further analyses. A summary of the data generation steps is shown in Figure S3.

2.4 Population genetic structure and genetic diversity

We assessed population structure with pairwise pool-F_{ST} and principal components analysis

(PCA). We computed the pool-F_{ST} ($\hat{F}_{ST}^{pool}$) statistic using the R package *poolfstat* (Hivert,

Leblois, Petit, Gautier, & Vitalis, 2018), which is equivalent to the Weir & Cockerham (1984)

F_{ST} and accounts for random chromosome sampling in pool-seq.

To evaluate whether patterns of population structure are better explained by neutral of

selective processes, we separately performed PCA on two SNP subsets, one of undifferentiated

(assumed neutral) and the other of highly differentiated markers (outliers, assumed selective).

These marker sets were chosen based on the empirical distribution of allele frequencies and

standard deviation (SD) cut-off values (Figure S4) (see Supplementary Data for details). To

reduce redundancy and physical linkage among SNPs, we retained one marker every 1 kb in the

undifferentiated marker set, and one marker every 10 kb in the differentiated set, as it is expected

that linkage is more pronounced in regions under selection. PCA was performed using the R

package *prcomp.* In a pilot analysis one sample from the west of Ireland (WIE2) appeared as an

outlier (Figure S5). Since no biological reason can explain this behavior, this sample was

removed.

To examine the genome-wide variation of genetic diversity in each pool, we calculated

nucleotide diversity (π) per pool in 10 kb-sliding windows with a step size of 2 kb using

PoPoolation 1.2.2 (Kofler et al., 2011) (see Supplementary Data for details). Plotting and

statistical testing was performed using the R environment (R Core Development Team, 2021).

Additionally, we evaluated whether population structure followed an isolation-by-

distance pattern with a Mantel test and 9999 permutations using the R package *ade4* (Dray &

Dufour, 2007). We compared the linearized genetic distances, calculated with the formula

linearized-$\hat{F}_{ST}^{pool} = \frac{\hat{F}_{ST}^{pool}}{1-\hat{F}_{ST}^{pool}}$ (François Rousset, 1997), and the geographic distances, estimated as

the least-coast oceanic distance in kilometers (km) considering land as barrier using the R

package *CartDist* (R. R. E. Stanley & Jeffery, 2017).

2.5 Detection of loci under selection

We applied the effective coverage correction (n_{eff}) to raw read counts to account for random

variation of read coverage and chromosome sampling during pooling and sequencing (Bergland,

Behrman, O'Brien, Schmidt, & Petrov, 2014; Feder, Petrov, & Bergland, 2012; Kolaczkowski,

Kern, Holloway, & Begun, 2011) (refer to Supplementary Data for details). Pool allele

frequencies were then computed using a python script.

To identify genomic regions with elevated differentiation respect to the genomic

background, we calculated the absolute delta allele frequency (dAF) per SNP between paired

contrasts of grouped pools, as *dAF = absolute(meanAF(group1) – meanAF(group2)*. The

contrasts evaluated were established based on geographic closeness, PCA clustering, and

biological knowledge (see details in Table S3). We also calculated the moving (rolling) average

of dAF values in 100-SNP windows to identify regions with consistent differentiation across

nearby markers, while ruling out single SNPs that could be influenced by random effects of pool-

seq. We further explored the allele frequency patterns of the most differentiated SNPs at each

locus and contrast for all the 12 pools. All the analyses were performed using R and plotting was

done with the R package *ggplot2* (Wickham, 2016).

2.6 Validation of informative markers for genetic stock assessment

To identify a reduced panel of highly informative SNPs for genetic stock identification and to

validate pool-seq findings, we obtained the genotypes of 160 individuals (20 fish each from eight

locations) in 100 SNPs (Table S4). The 100-SNPs panel consisted of 24 neutral markers and 76

putatively adaptive markers (see Supplementary Data for details). The final split of adaptive

markers was: North Sea (n=28), north-south break (n=13), west of Ireland (n=14), Alboran Sea

(n=13), southern Portugal (n=4), north Africa (n=4). Three to four individuals per location were

genotyped twice to assess genotyping error rate. DNA extraction and SNP genotyping were

undertaken by IdentiGEN, Ireland, using their IdentiSNP genotyping assay chemistry. The

protocol utilises target specific primers and universal hydrolysis probes. Following an end-point

PCR reaction, different genotypes are detected using a fluorescence reader.

Based on individual allele frequencies, we undertook a preliminary analysis of population

structure among the eight genotyped fish aggregations. It should be noted that sample sizes were

small and therefore the results of the population analyses should be viewed as preliminary until

further large-scale screening is undertaken. Only individuals and markers with >80% genotyping

success were retained. Deviations from Hardy–Weinberg equilibrium (HWE) and linkage

disequilibrium (LD) were assessed with *Genepop* 4.2 (Rousset, 2008). Six SNPs had indication

of deviation from HWE, two markers (12_3119866 and 17_972744) were not polymorphic and

one had evident scoring errors (24_5252083), thus these nine markers were excluded.

Microsatellite Analyzer (*MSA*) 4.05 was used to calculate pairwise F_{ST} estimates (Dieringer &

Schlötterer, 2003). In all cases with multiple tests, significance levels were adjusted using the

sequential Bonferroni technique (Rice, 1989). PCA was performed using the R function *prcomp*.

We estimated individual admixture coefficients for a given number of ancestral source

populations (*K*) using the *sNMF* algorithm (Frichot, Mathieu, Trouillon, Bouchard, & François,

2014a) of the R package *LEA* (Frichot, Francois, & Caye, 2017). We tested *K* = 1 to 5, with 10

repetitions and 200 iterations. The most likely *K* corresponds to the value where the cross-

entropy criterion (metric that evaluates the error of the ancestry prediction) plateaus or increases

(Frichot, Mathieu, Trouillon, Bouchard, & François, 2014b). We plotted the average admixture

proportions per population sample over a map using the R package *ggplot* (Wickham, 2016) and

ggOceanMaps (Vihtakari, 2020).

2.7 Characterization of a putative inversion on chromosome 21

To assess the genetic diversity and spatial distribution of haplotypes of the putative inversion on

chromosome (chr) 21, we extracted the individual genotypes of 12 diagnostic SNPs within the

inversion from the 100-SNP dataset (Figure S11). To identify the inversion genotype of each

individual, we performed a PCA with the R function *prcomp*. Individuals were assigned to each

haplotype group using the first two eigenvectors of the PCA and the k-means clustering

algorithm implemented in the R function *kmeans*. We calculated observed heterozygosity for

each of the PCA clusters, with the expectation that the middle cluster will have the highest

heterozygosity of all. These analyses and correspondent graphics were performed using R.

2.8 Genome-Environment Association

To identify which environmental variables are strongly correlated with adaptive genetic variation

and local adaptation, we tested genome-environment association using redundancy analysis

(RDA). We retrieved environmental data from *Bio-Oracle* v.2.1 (Assis et al., 2018; Tyberghein

et al., 2012) for each sampled location using the *R* package *sdmpredictors* (Bosch, 2020). Data

corresponded to mean bottom depth layers of minimum, mean, maximum, and range values of

sea water temperature (°C), sea water salinity (PSS), and dissolved oxygen concentration

(μmol/m^3) for a 14-year period (from 2000 to 2014) and a spatial resolution of 0.25 arcdegree.

This data was derived from a numerical model developed by the Global Ocean Biogeochemistry

non assimilative Hindcast (PISCES) (Copernicus Programme of the European Union, n.d.).

Environmental data per location were standardized to zero mean and unit variance, and

collinear variables were removed (see Supplementary Data for details). To run RDA, we used the

R package *vegan* (Dixon, 2003), the non-redundant environmental dataset, and the pool-allele

frequencies of SNPs in outlier loci found with genome scans. The significance of the RDA

model and environmental variables was assessed with an analysis of variance (ANOVA) using

1000 permutations.

2.9 Functional annotation of gene models

The gene models of the Atlantic horse mackerel genome were developed by Ensembl (Howe et

al., 2021) and become available in an Ensembl Rapid release since March 2021 (Ensembl,

2021b). However, at the time this research was conducted, the gene models lacked gene symbols

(names) and GO terms (only Ensembl gene IDs were available). Given that these annotations are

relevant for interpretation of results, we ran the functional gene annotation pipeline developed by

the National Bioinformatics Infrastructure Sweden (NBIS) (Binzer-Panchal, Dainat, & Soler,

2021) to retrieve this information (see Supplementary Data for details).

Additionally, for the top 2% most differentiated SNPs within each divergent genomic

region detected with genome scans, we annotated the closest overlapping gene (up to $\pm$ 40 kb)

and the variant effect prediction (e.g., missense, synonymous, upstream, downstream, intergenic)

using snpEff v.4.1 (Cingolani et al., 2012).

3. Results

3.1 Sequencing and variant calling

We generated pooled DNA whole-genome sequence data of 12 Atlantic horse mackerel aggregations sampled across the species' range in the east Atlantic and the western Mediterranean Sea, Alboran Sea (Figure 1A, Table 1). Temporal replicates of four locations were collected one year apart, so the temporal stability of their genetic composition could be investigated (west of Ireland WIE1-WIE2, North Sea NOS1-NOS2, northern Portugal NPT1-NPT2, and southern Portugal SPT1- SPT2). Our sequencing effort yielded 490-764 million high-quality reads per pool. 98.2-99.2% of reads aligned to the chromosome-level genome assembly of *T. trachurus* (Genner & Collins, 2022), and each pool had a mean depth of coverage between 25.7x and 46.3x (Table S5). After variant calling with GATK, a total of ~12.8 million biallelic SNPs passed quality filters and were used in the population analysis.

3.2 Population genetic structure

Pairwise pool-F_{ST} estimates indicated low genome-wide differentiation among horse mackerel populations (global mean pool-F_{ST}=0.007, range 0.001-0.012) (Figure 1B), but revealed four subtle population structure patterns that were also supported by genome scans. First, the western Mediterranean Sea (MED) differentiates from Atlantic populations (mean pool-F_{ST}=0.011). Second, Atlantic populations differ along a latitudinal cline that distinguishes "northern" or "southern" groups respect to a pronounced genetic break near Lisbon, Portugal (~38.7-39.0°N). Third, populations within the "northern" group (North Sea–NOS1, NOS2, west of Ireland–WIE1, WIE3, northern Spanish Shelf–NES, northern Portugal–NPT1) are genetically more similar to each other (mean pool-F_{ST}=0.004) than populations within the "southern" group (southern Portugal–SPT1 and north Africa–NAF) (mean pool-F_{ST}=0.006). Forth, the two samples from the North Sea (NOS1 and NOS2) show the highest genetic similarity (pool-F_{ST}=0.001) of all and are distinguished from other "northern" populations.

While genetic differences within the "northern" group are small, they are not negligeable (pool-F_{ST} range 0.001-0.006) (Figure S8). A significant Mantel test assessing the degree of correlation between genetic and geographic distances (H_o: no association) between pairs of populations, indicated that subtle genetic differences within the northern group follow an

isolation-by-distance (IBD) pattern, meaning that they are shaped by effective gene flow

between neighbouring areas (Mantel test p-value=0.015; linear regression $R.adjusted^2$=0.9, p-

value=1.1^{-13}) (Figure 1C).

We further examined whether population structure patterns were primarily driven by

neutral or selection processes by separately performing PCA for two SNP subsets, of

undifferentiated or highly differentiated markers. These marker sets were chosen based on the

standard deviation in allele frequencies across populations (see Methods for details, Figure S4).

The undifferentiated markers were considered "neutral" (61,543 SNPs) (Figure 1D, S8A), and

the highly differentiated markers were assumed under selection (818 SNPs) (Figure 1E, S8B). In

both PCA plots the separation between the "northern" and "southern" samples was evident,

although the composition of the clusters slightly varied depending on the marker set. In both

cases the "northern" group appeared as a tight cluster of genetically similar samples mainly from

the North Sea, west of Ireland, the northern Spanish shelf, and northern Portugal. In stark

contrast, the "southern" samples were dissimilar to each other. In the PCA with neutral markers,

the first two axis explained 30.6% of the genetic variance. The mid-range samples from the

Mediterranean Sea (MED) and southern Portugal (SPT1) were the most genetically distinct,

while the north Africa (NAF) sample was genetically similar to the "northern" group. The North

Sea samples tightly clustered with others in the "northern" group, being almost

undistinguishable. In contrast, in the PCA with selective makers, the first two axis explained

twice as much genetic variance (60.8%). The Mediterranean Sea sample was genetically closer

to the "northern" group, while the other two "southern" samples, from southern Portugal and

north Africa, were the most differentiated. Notably, the two North Sea samples formed a distinct

cluster separated from other "northern" samples.

Taken together, these results indicate that the separation between the westernmost part of

the Mediterranean Sea and Atlantic populations is largely driven by neutral processes, while the

latitudinal genetic cline and separation of North Sea samples could be the result of selective

processes.

3.3 Putative loci under selection

We performed whole genome scans based on the absolute difference in allele frequencies (dAF)

for paired contrasts to identify outlier loci respect to the genomic background (presumed under

selection). The contrasts were chosen based on PCA clustering patterns (Figure 1D, 1E), and the

outlier regions were identified using a Bonferroni Z-score threshold of significance. Our analysis

revealed a number of genomic regions with elevated differentiation for three contrasts: (i) the

Mediterranean Sea vs. others, (ii) "northern" vs. "southern" groups, and (iii) the North Sea vs.

others (Figure 2-4). A summary of the top 2% most differentiated SNPs per region, their closest

gene, gene function, and relative position to genes (e.g., missense, synonymous, upstream,

downstream, intergenic) inferred with snpEff are compiled in Table S6.

Two outlier loci distinguish the western Mediterranean from Atlantic populations

The contrast between Mediterranean Sea and other samples uncovered two distinctive regions at

chromosomes (chr) 5 and 21 (Figure 2). In this case the signals of divergence were not as evident

as in other contrasts given the overall higher genome-wide differentiation of this sample. Thus,

we focused on the peaks revealed by the rolling average of dAF values with consistent allele

frequency differences for the contrast of interest. We considered candidate genes those

overlapping or located in the vicinity of the top 2% most differentiated SNPs in each locus. In

the region on chr 5 (Figure 2B), the Mediterranean Sea sample had intermediate allele

frequencies across loci and higher nucleotide diversity, while other samples tended to be fixed

for one haplotype and showed lower nucleotide diversity. The top candidate gene in this region is

opn1mw4 (*green-sensitive opsin-4*). In the putative selection signal on chr 21, the Mediterranean

sample had one haplotype close to fixation, the opposite haplotype was in high frequency in the

North Sea, and intermediate frequencies were prevalent elsewhere (Figure 2C). This region

harbours a single gene, *taar7a* (*trace amine-associated receptor 7A*). The current knowledge of

potential function of candidate genes is presented in Table S6.

A presumed 9.9 Mb inversion distinguishes populations along a latitudinal cline

With the "northern" vs. "southern" contrast, we discovered that a single large locus on chr 21

underpins the latitudinal cline with a genetic break along mid Portugal (Figure 3). This locus

consists of a large block of SNPs with elevated dAF spanning 9.9 Mb (Figure 3A-B). The large

size and abrupt drop in allele frequency differences towards the edges are characteristic of

structural variants with suppressed recombination (e.g., inversions) that do not represent recent
selective sweeps (Han et al. 2020). An analysis of the pool-allele frequencies of the top 2% most
differentiated SNPs at this locus (Figure 3C) revealed striking allele frequency differences
between the "northern" group (including the Mediterranean Sea) and north Africa, while
intermediate allele frequencies were prevalent in southern Portugal.
To further characterize this structural variant, we leveraged genotype data of 160
individuals for 12 SNPs within the region. As expected for a chromosomal inversion, in a PCA
plot individuals separated into three clusters depending on their genotype (Figure 3D), and
individuals in the intermediate cluster exhibited the highest average observed heterozygosity
compared with the other two (Figure 3E). Overall, the individual genotype data supported the
observations made with pool-seq data regarding the spatial distribution of inversion haplotypes,
although it offered greater resolution. For instance, it revealed that the "northern" haplotype
dominates in the North Sea, west of Ireland, northern Spanish shelf, and northern Portugal, but
also occurs at an intermediate frequency in southern Portugal. The "southern" haplotype is
predominant in north Africa, and common in southern Portugal (Figure 1F). This putative
inversion contains about 1,077 genes, making it difficult to infer which particular genes are
under selection.
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A genetically distinct population in the North Sea
The contrast between the North Sea and other samples revealed seven genomic regions with
elevated differentiation on chr 1, 4, 7, 11, 20, and 21 (Figure 4A). Further examination of allele
frequencies of top SNPs per locus indicated that the North Sea temporal replicates were very
genetically similar, and that their allele frequency patterns were distinctive of this location
(Figure 4B-E, Figure S9). For instance, in the selection signals on chr 1 and 21, North Sea
samples had variant alleles close to fixation and a slight reduction in nucleotide diversity (π),
suggesting the possibility of a soft selective sweep (Figure 4B, 4C, 4E). At the outlier loci on chr
4, 11, and 20 (Figure S9B, S9D, and Figure 4E, respectively), North Sea samples had
predominantly intermediate allele frequencies, with similar (chr 4 and 11, Figure S9B, S9D) or
higher (chr 20, Figure 4D) nucleotide diversity with respect to other pools. Additional divergent
regions on chr 7 and 11 showed a similar but noisier pattern, presumably due to the presence of
complex structural variants at these loci (Figure S9C, S9D).

The most striking signals were located on chr 1, followed by those on chr 20 and 21

(Figure 4). The signal on chr 1 encompasses two regions (chr1:17400000-17600000,

chr1:22020000-22100000). In the first region (Figure 4B), the most differentiated SNPs are in

the vicinity of the gene *gpr83* (G-protein coupled receptor 83). In the second region, the top

candidate gene is *sgms2* (*sphingomyelin synthase 2*) (Figure 4C). The signals on chr 20 (Figure

4D) and chr 21 (Figure 4E) contain a single gene each, *ncoa2* (*nuclear receptor coactivator 2*)

and *taar7a* (*trace amine-associated receptor 7A*). Note that the signal in chr 21 is the same as

one previously described distinctive in the Mediterranean. Additional signals were found on chr

4, 7 and 11 (Figure S9), however they were not as clear as the ones just described (e.g., smaller

allele frequency differences or inconsistency allele pattern). Therefore, it was difficult to identify

candidate genes for these regions. The current knowledge concerning the function of candidate

genes in the differentiated genomic regions is summarized in the Table S6.

3.4 Genome-Environment Association

To identify the environmental variables likely driving adaptive genetic variation, we retrieved

sea water environmental variables (temperature, dissolved oxygen, and salinity) and performed

genome–environment association (GEA) based on redundancy analysis (RDA). A total of

210,568 SNPs within the most divergent genomic regions in the North Sea, Mediterranean Sea,

and northern vs. southern contrasts were used in this analysis.

Mean sea water dissolved oxygen concentration has the strongest association with genetic

variation ($R^2 = 0.83$), followed by mean sea water temperature ($R^2 = 0.71$) (Figure 5) and, to a

lesser extent, with salinity ($R^2 = 0.34$). Notably, the latitudinal gradient shows an inverse

relationship with oxygen content, with the North Sea being the location with the highest oxygen

levels (~270-300 μmol/m^3), almost three times the oxygen content in north Africa (~100

μmol/m^3). The two southernmost populations studied (in north Africa and southern Portugal)—

which showed a high and intermediate frequency of the "southern" inversion haplotype,

respectively—are characterized by higher mean temperatures (18°C), with respect to locations

north of Lisbon, in which the "northern" haplotype dominates. The North Sea has the coldest

temperatures of all locations (< 10°C). In terms of salinity, as expected, the Mediterranean Sea

has the highest salinity followed by southern Portugal, while the lowest salinity is observed in

the North Sea. That being said, the salinity range between the northern- and southernmost

locations here analysed is minimal, between 35.0 and 36.5 PSS, which is considered a fully

marine environment.

3.5 Validation of informative markers for genetic stock identification

To confirm the selection signals found with pool-seq data and develop a tool for genetic stock

identification, we genotyped 160 individuals in 100 SNPs. Of these, 24 markers were neutral and

76 were among the most differentiated across contrasts (Figure S11). We found a strong

correlation between allele frequencies calculated from individual genotypes and pool-seq data

(mean $R^2 = 0.9 \pm 0.1$), supporting the findings of the pool-seq analysis (Figure S12). Of the 76

outlier loci, 72 had a genotyping success >80% (Table S7). After applying quality filters, the

final dataset had 63 SNPs and 157 out of 160 individuals had a genotyping success >80%

(dataset called the 63-SNP panel).

To minimise marker redundancy in the 63-SNP panel, we performed a linkage

disequilibrium (LD) analysis for all loci and samples. As expected, significant LD was found

between a number of SNPs located in close proximity on the same chromosome (Table S7).

Though LD was not statistically significant in some cases (e.g., SNPs on chr 5), these were

considered to be linked due to physical closeness. To identify the most informative SNPs for

discriminating the samples while reducing LD, we analyzed the F_{ST} by marker and by population

(Figure S13) for each genomic region. We retained the SNP with the highest average F_{ST} per

linkage group (assumed to be the most informative), yielding a 17 SNP dataset comprising 155

out of 160 individuals with a genotyping success >80% (dataset called the 17-SNP panel).

Further analyses were conducted with both the 63-SNP and the 17-SNP datasets (individual

genotypes are shown in Figure S14).

An examination of pairwise F_{ST} values indicated lack of significant genetic differences

between the North Sea temporal replicates or between the two west of Ireland samples (Table

S8). There was also no significant genetic differentiation between the individuals from the

northern Spanish shelf, northern Portugal, and the two samples from the west of Ireland (Table

S8).

To assess whether the same population structure is achieved with the 63- and 17-SNP

panels, we performed separate PCA and admixture analyses. We observed similar clustering

patterns with both panels (Figure 6A, 6B), consisting of four main groups: (1) North Sea (NOS1,

NOS2); (2) west of Ireland (WIE1, WIE2), northern Spanish shelf (NES), and northern Portugal

(NPT1); (3) southern Portugal (SPT1); and (4) north Africa (NAF). While the 63-SNP panel

provided better discrimination between southern Portugal (SPT1) and north Africa (NAF), both

marker sets separated most individuals from the North Sea, as well as the "northern" and

"southern" groups. It should be noted that some individuals clustered in different groups from

the ones expected, suggesting that gene flow occurs among geographic regions.

An individual admixture analysis with both datasets supported the same four major

genetic groups (Figure 6C, 6D) as the ones identified with PCA, showing various levels of

mixing among them. For example, a few individuals in the North Sea (n = 3-4) had a high

probability of originating from the western group (first-generation migrants). Similarly, southern

Portugal had a number of individuals that appear to originate from the western group (n = 3) or

from the African group (n = 3). In all four groups some individuals showed admixed ancestry,

indicating that they are probably F1-hybrids or backcrosses between local and migrants. Overall,

these results indicate that gene flow occurs more often between neighboring geographic areas,

following an isolation-by-distance pattern (e.g., between the North Sea and the west of Ireland;

between the northern Spanish shelf, northern Portugal and southern of Portugal; and between

southern Portugal and north Africa).

4. Discussion

This study represents the most comprehensive assessment of the genomic variation of Atlantic

horse mackerel populations to date (Figure 1). The combination of extensive geographic

sampling and genomic screening derived from whole-genome sequencing, provided a powerful

dataset that allowed us to discover genomic regions underlying population structure and local

adaptation. Moreover, we developed a genetic tool that can help distinguish some of the main

biological groups found.

Population structure and temporal stability of genetic groups

Our genomic analysis indicated that very few genetic differences exist between horse mackerel

populations inhabiting the vast area from the North Sea to north Africa (Figure 1A), more than

4,000 km of coastline (global mean pool-F_{ST}=0.007, range: 0.001-0.012). This result is in

agreement with previous studies using a few genetic markers (Cimmaruta et al., 2008; Comesaña

et al., 2008; Healey et al., 2020; Kasapidis & Magoulas, 2008), and implies that gene flow is

pervasive. Low genetic differentiation can be explained by minimal barriers to gene flow and

large population sizes, commonly observed in migratory marine fish. We infer that in the horse

mackerel, since the pelagic larval stage lasts about one month (Rusell, 1976), gene flow could

predominantly occur between neighbouring areas through passive transport of eggs and larvae by

ocean currents. Moreover, the large population sizes of this species imply a minimal role of

genetic drift in shaping patterns of genetic diversity.

Despite low differentiation at neutral loci, we uncovered patterns of population structure

(Figure 1B, D, E) mainly at loci putatively under selection (Figures 2-4). Below, we will touch

on the summary statistics results, and in the next section we will discuss the genome scan results.

The greatest genetic differences found occurred primarily at neutral markers between the

Alboran Sea (the westernmost part of the Mediterranean) and the Atlantic. This separation was

already proposed in earlier studies using body morphometrics, otolith shape, and parasitofauna

(Abaunza et al., 2008). However, this is the first time that genetic data supports it, as previous

studies using a few markers were inconclusive (Cimmaruta et al., 2008; Comesaña et al., 2008;

Healey et al., 2020; Kasapidis & Magoulas, 2008). The Mediterranean-Atlantic divide has also

been described in several marine species but not in all (Patarnello, Volckaert, & Castilho, 2007),

likely because of differences in species' life history or low resolution of the genetic markers

used. Previous research based on parasite composition indicated that the Alboran Sea is a mixing

area of horse mackerel from the Mediterranean and the Atlantic (Abaunza et al., 2008; Mattiucci

et al., 2008). If this were the case, we would expect to see both inversion haplotypes at this

location, but what we observed is that the northern haplotype prevails. This, in addition to the

fact that none of the individuals collected were actively spawning (Table S1), let us infer that our

sample is probably not representative of this area, and likely comes from a region with colder

waters (e.g. Gulf of Lyon, Balearic Sea) (Figure S10). A possible neutral mechanism for reduced

gene flow between Atlantic and western Mediterranean populations is the presence of retentive

currents in the Almeria-Oran front, which have been proposed to act as barriers for gene flow for

various marine species (Patarnello et al., 2007).

Our genomic data also revealed an undescribed latitudinal genetic break in mid Portugal,

near Lisbon (~38.7-39.0°N), distinguishing populations from the north or south of this area. This

genetic separation was noticeable in pairwise-F_{ST} estimates, but it was more evident in the PCA

based on outlier SNPs, suggesting a putative role of natural selection. Interestingly, the inferred

location of the north-south genetic break coincides with a major biogeographical transition zone

between temperate and subtropical waters off the coast of Portugal, which have differential

species composition (Cunha, 2001; Santos et al., 2007). Moreover, a recent genetic study in the

boarfish (*Capros aper*), a pelagic fish with similar distribution and life history as the horse

mackerel, reported a comparable latitudinal pattern (Farrell, Carlsson, & Carlsson, 2016). Thus,

it is possible that the environmental transition zone in Portugal is a major driver of population

structuring of many marine species inhabiting this area, including the horse mackerel.

Within the "northern" group—North Sea, west of Ireland, the northern Spanish shelf

(Bay of Biscay), and northern Portugal—genetic differences are small at both, neutral and

selective markers (Figure 1B, 1D). Such genetic variability resembles an isolation-by-distance

pattern (Mantel test *p*-value=0.015), which was also observed in a previous study with allozymes

(Cimmaruta et al., 2008). Therefore, genetic differences between populations within the

"northern" group seem to be largely determined by the effective gene flow via passive larval

dispersal by ocean currents. The North Sea samples though, were slightly more differentiated at

outlier SNPs (Figure 1E), suggesting a putative role of natural selection.

In contrast, populations within the "southern" group—inhabiting waters in southern

Portugal and north Africa, Mauritania—are genetically more differentiated between them and

from all other populations. Interestingly, the population in the south of Portugal was different at

both, neutral and selective markers, whereas the population from north Africa was more

genetically distinct at outlier markers. We will discuss these sub-structuring patterns in more

detail in the genome scan section.

In summary, our genomic data indicates five possible genetic groups: (1) the western

Mediterranean Sea; (2) the North Sea; (3) the west of Ireland, northern Spanish shelf (Bay of

Biscay), and northern Portugal; (4) southern Portugal; and (5) north Africa, Mauritania. These

groups seem to be temporally stable, as indicated by the similar allele frequencies at outlier loci

observed between temporal replicates (west of Ireland WIE1-WIE2, North Sea NOS1-NOS2,

northern Portugal NPT1-NPT2, and southern Portugal SPT1- SPT2) (Figures 2-4).

Genomic regions and environmental factors involved in local adaptation

Genome scans uncovered a number of regions with elevated differentiation that support the

subtle patterns of population sub-structuring identified with summary statistics. The informative

contrasts were: (i) the Mediterranean Sea vs. others, (ii) "northern" vs. "southern" groups, (iii)

the North Sea vs. others. The divergent genomic regions varied in size (70-600 kb, and 9.9 Mb

for a structural variant) and harboured numerous candidate genes.

The test for genome-environment association (GEA) indicated that genetic differences at

putatively adaptive loci are primarily driven by differences in sea water oxygen content and

temperature between locations (R^2=0.83 and R^2=0.71, respectively) (Figure 5). Salinity seemed

to be less important, reflecting the minute differences in salinity between areas (R^2=0.34, range

35-36.5 PSS). Notably, there was an inverse relationship between sea water oxygen, temperature,

and the latitudinal cline, where the North Sea had the highest oxygen concentration (~270-300

μmol/m^3), almost three times the oxygen content in north Africa (~100 μmol/m^3). Moreover, the

North Sea also exhibited the coldest sea water temperature of all locations analyzed (< 10°C),

strikingly different from the two southernmost populations sampled (north Africa and south

Portugal) (18°C).

We found that two outlier loci distinguish the western Mediterranean Sea, one on chr 5

and the other on chr 21 (Figure 2, Figure S10). The signal on chr 21 harbours a single gene,

taar7a (*trace amine-associated receptor 7A*), which encodes a receptor involved in olfactory

sensing of amines (Hashiguchi & Nishida, 2007). The signal on chr 5 includes three genes that

encode proteins with crucial roles in synapsis (*cadps, synpr, slc6a13*), and five in tandem copies

of the green-sensitive opsin-4 (*opn1mw4*) gene. This gene organization (*opn1mw4* copies flanked

by *synpr* and *slc6a13*) is highly conserved among several Clupeocephala species (e.g., Atlantic

herring, zebrafish, medaka, stickleback, (Lin, Wang, Li, & Wang, 2017; Nakamura et al., 2013)),

suggesting that it has been maintained by natural selection.

The top candidate gene at this locus is *opn1mw4*, a paralog of the *opn1mw* (RH2) gene,

which encodes cone photopigments that are essential for vision of blue-green light (medium

wavelength). This gene contains two missense mutations (c.850G>A and c.670G>A, Figure S10

A, B), one more divergent than the other (dAF=0.53 and dAF=0.40, respectively). The most

divergent variant results in an amino acid change from alanine to tryptophan at residue 284

(p.Ala284Thr), and the less divergent variant, from valine to isoleucine at residue 224

(p.Val224Ile). While all four amino acids are hydrophobic, tryptophan is more hydrophobic and

larger in volume than alanine, and the equivalent occurs between isoleucine respect to valine,

suggesting that these amino acid changes could potentially result in structural protein

modifications (Behmard, Abdolmaleki, & Asadabadi, 2012). Therefore, it is possible that the

missense mutations in *opn1mw4* are causal, as they could potentially affect the structure of the

photopigments and, thus, generate a shift in spectral sensitivity.

A change in visual sensitivity could be an adaptive response to contrasting light

environments between the Mediterranean and the Atlantic. In clear waters the blue-green light

spectrum dominates, especially as depth increases, while in shallower or turbid waters red-

shifted light conditions prevail. Comparative genomic studies indicate that fish living in marine

habitats dominated by green-blue light (e.g., in the deep sea, the pelagic open ocean, or in coral

reefs at night) commonly show gene copy expansions of *opn1mw* (RH2) respect to fish living in

shallower (<30 m) or more turbid waters (Lin et al., 2017; Musilova & Cortesi, 2021). Indeed,

the horse mackerel has five tandem copies of green-light opsins, which might relate to their

preferred habitat at deeper waters (100–200 m depth) (Brunel et al., 2016). Sea water turbidity in

the Mediterranean Sea is generally lower than in the Atlantic Ocean (Figure S10C), suggesting

that more green light might be transmitted to deeper layers of the water column.

Taken together, the probable adaptive significance of the missense mutations in *opn1mw4*

is that they might be involved in spectral fine-tuning of green-opsins to downwelling light

detection under contrasting light conditions between the Mediterranean Sea and the Atlantic. The

former likely dominated by blue-green light conditions and the latter, by a red-shifted light

environment. This visual adaptation could confer survival advantages related to feeding,

recognition of conspecifics, and escape from predators.

We discovered that the latitudinal genetic cline is underlined by a large (9.9 Mb) putative
chromosomal inversion on chr 21 (Figure 3). The northern haplotype shows high frequency in
the "northern" group and in the western Mediterranean Sea, while the alternative haplotype
dominates in north Africa, and intermediate frequencies of the two haplotypes occur in the south
of Portugal. This chromosomal inversion harbours thousands of genes, for which it is impossible
to infer its role without further studies. However, considering the strong genetic differentiation it
drives along an important oceanographic transition zone, we infer this inversion may play a
fundamental role in local adaptation to contrasting thermal regimes: a temperate (cooler) climate
in the north, tropical (warmer) in the south, and variable/intermediate at the transition zone in
mid Portugal. A possible explanation that the northern haplotype is prevalent in the western
Mediterranean Sea is that this sample might actually spawn in a different location with cooler
waters than where it was collected (near the Gibraltar strait, warmer temperatures similar to
those in southern Portugal, ~16 °C). Within the Mediterranean Sea, comparable temperatures to
those in the northern region (~11-12 °C) occur in the winter season in coastal waters of the
Balearic Sea near Catalonia (Spain), the Gulf of Lyon (France), and the Ligurian Sea (northeast
Italy) (Pastor, Valiente, & Khodayar, 2020, see Figure 3). Indeed, previous studies have reported
spawning events of horse mackerel near Catalonia during the winter months (Andreu &
Rodriguez-Roda, 1953; Planas & Vives, 1953). Given that the individuals collected in the
Alboran Sea were not actively spawning, we infer that they likely come from a region further
north, near the Gulf of Lyon.
636
We uncovered a genetically distinct population in the North Sea. We reached this
conclusion because seven genomic regions clearly distinguish samples from this location, and
the temporal replicates showed nearly identical allele frequencies at outlier loci. This observation
adds to previous morphometric and parasite data suggesting that horse mackerel from the North
Sea differs from nearby Atlantic populations (Abaunza et al., 2008).
Some of the top candidate genes likely involved in local adaptation to the North Sea are
gpr83, *sgms2*, *ncoa2*, and *taar7a*. *gpr83* (G-protein coupled receptor 83) encodes a receptor that
play an important role in the regulation of energy metabolism, feeding, reward pathway, and
stress/anxiety responses in mice, for which it has been linked with the control of body weight
(Gomes et al., 2016; Lueptow, Devi, & Fakira, 2018). *ncoa2* (*nuclear receptor coactivator 2*)

encodes a transcriptional coactivator for steroid receptors that is presumably involved in glucose

metabolism regulation (Bateman et al., 2021). Previous experimental studies indicate that fish

adapted to cold climates often have higher metabolic rates than those adapted to warm climates

(Wang, Tan, Jiao, You, & Zhang, 2014; White, Alton, & Frappell, 2012). Therefore, we infer

that selection is likely favoring alleles that result in increased energy metabolism required for

survival in the colder environment of the North Sea.

sgms2 (*sphingomyelin synthase 2*) encodes a protein that is involved in the synthesis of

sphingomyelin, a major component of cell and Golgi apparatus membrane. Previous studies in

fish indicate that this protein is crucial to maintain cell membrane structure and fluidity at low

temperatures (Wang et al., 2014; Windisch, Kathöver, Pörtner, Frickenhaus, & Lucassen, 2011).

Thus, we infer that genetic variants in this gene could facilitate cold tolerance in the North Sea.

taar7a (*trace amine-associated receptor 7A*) codifies an olfactory receptor specific for

sensing amines in vertebrates (Hashiguchi & Nishida, 2007; Hussain et al., 2009; Tessarolo et

al., 2014; Yamamoto et al., 2010). Amines are behaviorally relevant odorants that are proposed

to play a critical role in intra- and interspecific communication, for example, in sexual attraction

or in avoidance of predators or rotting food (Dewan, 2021). A previous study on two goatfish

species with contrasting bottom habitat preferences (*Mullus surmuletus* lives on sandy and rocky

bottoms, and *Mullus barbatus* lives at deeper waters on muddy bottoms) reported significant

differences in the morphology of chemoreceptors. Such differences are proposed to be associated

with increased sensitivity to chemical stimuli in the species living in muddy and deeper water,

which has reduced visual capabilities. Among the locations in the east Atlantic included in this

study, the North Sea has the highest water turbidity of all (Shi & Wang, 2010)(Figure S10).

Thus, we infer that natural selection is likely favoring alleles in the gene *taar7a* that confer an

enhanced sense of smell under the reduced visibility in the North Sea.

Other candidate genes distinctive of the North Sea might play critical roles in

reproduction (spermatogenesis-*ankrd49*, gametogenesis-*spin1*, sex determination-*hsd11b1l*,

fertilization-*cdk2ap2*); development of bone (*papss1*), skeletal muscle (*spin1*) and skin (*ptk6*);

cell processes (*aip*, apoptosis-*nol4-nol4l*, nutrient-dependant growth-*srms*); energy metabolism

and body weight (glucose-*eno2-ncoa2*, lipids-*cyp2u1*, nitrogen-*slc38a3*, weight maintenance-

hadh); nervous and olfactory system (neurotransmitter release, *doc2b*; synaptic transmission,

slc38a3); immune response (*tmem134, serping1, clec4e*); and osmolarity regulation at low

temperatures (*hsd11b1l*). The complete list of candidate genes and their putative function are

presented in Table S6.

Interestingly, the number of divergent loci and their degree of differentiation in the

Atlantic horse mackerel are relatively small compared with those in the Atlantic herring (Han et

al., 2020), but is rather intermediate to the Atlantic herring and the European eel, which in fact

constitutes a single panmictic population (Enbody et al., 2021). We propose that the most

important explanation for the differences in genetic structuring between these three species is the

striking diversity of ecological conditions during spawning and early development, the most

sensitive period of life for a fish and a period characterized by high mortality (Dahlke, Wohlrab,

Butzin, & Pörtner, 2020). The Atlantic herring is a demersal spawner that breeds close to the

coast in areas with marked environmental differences as regards to temperature, salinity, depth

and biotic conditions (plankton production, predators, etc.). It also presumably shows homing

behaviour, that is, individuals return for spawning to the locations where they were hatched. In

contrast, the Atlantic horse mackerel is a benthopelagic fish that spawns at deeper waters, near

the shelf edge (100-200 m deep) at not well-defined spawning areas, implying that the range of

environmental conditions in which breeding occurs are comparatively less diverse. Finally, to the

best of our knowledge, all European eels spawn in the Sargasso Sea under similar environmental

conditions.

In summary, we revealed a number of highly divergent genomic regions encompassing

genes that may be involved in adaptation to local environmental conditions in the North Sea, the

western Mediterranean Sea, and along a climate-related latitudinal cline in the Atlantic. In

addition, our results highlight the importance of chromosomal rearrangements for local

adaptation with gene flow.

Implications for fisheries management

The genetic-based groups identified here are largely in agreement with the current horse

mackerel stock divisions, informed by the results of the 2000-2003 HOMSIR project (Abaunza

et al., 2008). However, our data do not support the current definition of the "southern stock" in

Portuguese waters (Figure S1), and of the southern boundary of the "western stock".

Our genomic data indicates that the "southern stock" actually consists of two biological

units (Figures 1, 2, 5), not one as previously assumed. Samples from northern Portugal (north of

Lisbon, ~38.7-39.0°N) appear to be genetically closer to the "western stock", while samples

from southern Portugal (south of Lisbon) form their own group but are genetically closer to the

samples from the "Saharo-Mauritanian stock", in north Africa. To confirm these findings and

assess the spatial and temporal trends of mixing between these areas, further studies are required,

including a finer geographic sampling and screening of informative genetic variants in a large

number of individuals throughout this area.

We did not find significant genetic differences between northern Portugal (currently

considered part of the "southern stock"), northern Spanish shelf (Bay of Biscay), and the west of

Ireland, implying that the southern boundary of the "western stock" could possibly be extended

down to the north of Portugal. However, the minute genetic differentiation does not exclude the

possibility that isolation in an ecologically-relevant timescale of interest for fisheries

management might occur (Hauser & Carvalho, 2008b).

Finally, our genomic data support the consideration of the Mediterranean Sea as a

separate stock, as proposed by the HOMSIR project (Abaunza et al., 2008). While a single

sample from the westernmost part of the Mediterranean was studied, its genetic distinctiveness

suffices to infer that the Mediterranean horse mackerel likely constitutes a separate population

from those in the Atlantic. Wide-scale sampling within the Mediterranean Sea is required to

further explore the population structure in this region.

This study disclosed a number of genetic markers (SNPs) that can be used as a genetic

tool for fisheries stock assessment. A panel of only 63 markers suffice to identify the main

genetic subdivisions. In fact, using a reduced panel of only 17 markers, it is possible to

differentiate individuals collected in "the North Sea stock" and "western stock", and from the

"northern" (North Sea, west of Ireland, northern Spanish shelf, northern Portugal, and the

Mediterranean) and the "southern" (southern Portugal, north Africa) groups with respect to the

genetic break along Lisbon. Thus, these markers can help, for instance, to elucidate the extent of

mixing between the Western and North Sea stocks in the English Channel (ICES Divisions 7.e

and 7.d) and in ICES area 4.a in the northern North Sea.

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Data Accessibility

Sequence data are available in NCBI under the bioproject XX, Genbank accession XX

Custom scripts are available in the Github repository: https://github.com/LeifAnderssonLab/

2022_Horse_mackerel_popgen.

Author contributions. EDF and LA conceived the study and obtained funds. APF-P performed the bioinformatic data analysis. EDF performed the genotype-based data analysis. MEP contributed to data analysis. CGS contributed to laboratory work. APF-P, EDF, and LA wrote the paper with contributions from all other authors. All authors approved the manuscript before submission.

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Figure 1. Population structure of the Atlantic horse mackerel. (**A**) Map depicting the 12 sampling sites in the east Atlantic Ocean. ICES fishing divisions are denoted in light blue, same as their alphanumerical code. The approximate location of a biogeographical transition zone north of Lisbon, in Portugal (~38.7-39.0°N), is denoted with a horizontal dashed line (Cunha, 2001; Santos et al., 2007). In all plots, each dot represents a sampling location and its color indicates the corresponding ICES stock division (ICES, 2005) after the HOMSIR project (Abaunza et al., 2008). (**B**) Heatmap plot representing pairwise pool-F_{ST} values based on ~12 million SNPs. (**C**) Isolation by distance (IBD) plot illustrating the relationship between genetic distances (linearized pairwise pool-F_{ST}) and geographic distances in km for the populations in the "northern" group. This analysis was performed excluding the Mediterranean sample (MED). The dashed line indicates the linear relationship between variables. (D-E) Principal component analysis (PCA) plot based on (**D**) undifferentiated (61,543 SNPs) and (**E**) highly differentiated (818 SNPs) markers. The first two axes are shown. Inset bar plots in each PCA plot shows the percentage (%) of genetic variance explained by the first nine principal components (PC). Note that samples WIE3, NPT2, and SPT2 were excluded from analyses B to E as they behaved as outliers likely due to technical reasons. Sample name abbreviations as in Table 1.

Figure 2. Two selective regions distinctive of the western Mediterranean Sea. (**A**) Manhattan plot depicting the dAF per SNP across the genome for the contrast between the Alboran Sea vs. all other samples. Each dot is a single SNP, and an alternating color (gray, blue) was used to differentiate SNPs in consecutive chromosomes. The line in black towards the bottom of the plot is the rolling average of dAF over 100 SNPs. Regions of interest are indicated with an arrow. Close-up plot for the signals in (**B**) chr 5 and (**C**) chr 21. Each close-up plot consists of four tracks: the first, is a representation of the gene models; the second, shows the dAF of SNPs, denoting in black the top 2% most differentiated SNPs; the third, shows the pool minor allele frequency of the top 2% SNPs in a heatmap plot, in which each row is one pool sample and each column is one SNP; and the fourth, illustrates the percentage of nucleotide diversity, π (%) along the region with a separate line for each pool sample, colored based on the ICES stock divisions. Sample name abbreviations as in Table 1.

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Figure 3. Putative chromosomal inversion on chromosome 21 that underlies a steep north-south genetic break
in Atlantic horse mackerel. (**A**) Manhattan plot representing the absolute difference in pool-allele frequencies
(dAF) of each SNP along the genome for the contrast between north vs. south groups (see Methods for details). The
x-axis shows the genomic position of each SNP and the y-axis, its dAF value. SNPs in consecutive chromosomes are
distinguished with alternating colors, gray and blue. The black line towards the bottom of the plot corresponds to the
100 SNPs-rolling average of dAF values. (**B**) Close-up Manhattan plot for chromosome 21. (**C**) Pool-allele
frequency per population sample (rows) of the top 2% markers (columns) within the putative inversion. (**D-F**)
Haplotype frequency based on the genotype of 15 SNPs screened in 160 individuals. (**D**) PCA plot showing
individual clustering respect to the inversion genotype. (**E**) Individual observed heterozygosity in each PCA cluster.
Clusters "1" and "3" correspond to homozygous "southern" and "northern" individuals, respectively, while cluster
"2" are heterozygous individuals, as indicated by their higher average observed heterozygosity respect to the other
two clusters. (**F**) Map showing the geographic distribution of inversion haplotypes across sampling locations. Pie
charts represent the frequency of haplotypes "North" or "South" per location. Sample name abbreviations as in Table
1.

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Figure 4. Genomic regions characteristic of the North Sea populations. (A) Manhattan plot representing the dAF
of each SNP along the genome for the contrast between the North Sea vs. all other samples. Each dot is a single SNP.
The black line towards the bottom is the rolling mean of dAF over 100 SNPs. Close-up plots of the four most
divergent regions in **(B-C)** chr 1, **(D)** chr 20, and **(E)** chr 21 (highlighted with arrows). Plots of the other
differentiated regions on chr 4, 7 and 11 are shown in the Supplementary Figure S9 (highlighted with an asterisk).
Each close-up plot consists of four sections (from top to bottom). The first section illustrates the gene models. The
second section corresponds to the dAF of SNPs, in which the top 2% of markers are denoted in black. The horizontal
red line indicates the Bonferroni Z-score threshold of significance. The third section is a heatmap plot depicting the
pool-allele frequency of the top 2% SNPs, where the rows are pool samples and the columns are SNPs. The fourth
section is the nucleotide diversity (π) for each pool sample calculated over 10 Kbp sliding windows with a step size
of 2 Kbp. The color of each line indicates the designated ICES stock division of each pool prior to the HOMSIR
project (Abaunza et al., 2008). Sample name abbreviations as in Table 1.

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Figure 5. Association between genetic variants and environmental variables. Analysis based on the pool-allele
frequencies of 210,568 SNPs within the most divergent genomic regions in the North Sea and the Mediterranean Sea,
and underlying the north-south genetic break. (**A**) Redundancy Analysis (RDA) plot showing the association
between pool-allele frequencies of SNPs and three uncorrelated environmental variables: Tmean, mean sea water
temperature (°C); O2mean, mean dissolved oxygen concentration ($\mu mol/m^3$); and Smean, mean sea water salinity
(PSS). Blue arrows represent the loadings of the environmental variables that contribute most to the first two
axes. Each dot represents a single pool sample, and its color indicates its ICES stock division. (**B-D**) Scatter plots
showing the linear relationship (dotted line) and 95% confidence interval between average pool-allele frequencies
and (**B**) sea water temperature, (**C**) mean dissolved oxygen concentration, and (**D**) mean sea water salinity (PSS).
Sample name abbreviations, NOS1-2: southern North Sea, WIE1-3: western Ireland, NPT1-2: northern Portugal,
SPT1-2: southern Portugal, MED: Mediterranean, NAF: North Africa.
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Figure 6. Population structure based on individual SNP genotyping. The two SNP datasets consisted on the genotype of 157 and 155 individuals screened using 63 and 17 markers, respectively. Principal components analysis plot based on (**A**) the 63-SNP dataset or the (**B**) the 17-SNP dataset. Only the first two axes are shown. Each dot corresponds to an individual, the dot color indicates sampling location and the point shape, the designated stock based on the ICES stock divisions. Inset bar plots show the percentage (%) of the genetic variance explained by each of the first nine principal components (PCs). (**C-D**) Admixture bar plot (left) and map showing the mean ancestry proportion per pool sample (right) for a number of ancestral populations (K), here equals to four, based on (**C**) the 63-SNP dataset, or the (**D**) the 17-SNP dataset. In the admixture bar plot, each row is one individual, and the different colors per individual represent the probability of ancestry for a given K. Individuals collected in the same site are grouped as a block. Sample name abbreviations as in Table 1.

1142 **Table 1.** Collection details of the 12 Atlantic horse mackerel samples included in this study.

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Pool ID	Stock [*]	Area	Year	Latitude	Longitud	Sample size
WIE1	Western	West of Ireland	2016	54.42	-10.62	51
WIE2	Western	Southwest of Ireland	2016	51.35	-10.98	44
WIE3	Western	West of Ireland	2017	52.76	-11.00	62
NOS1	North Sea	Southern North Sea	2016	54.15	3.30	96
NOS2	North Sea	Southern North Sea	2017	54.07	2.85	70
NPT1	Southern	Northern Portugal	2016	39.83	-9.20	64
SPT1	Southern	Southern Portugal	2016	37.26	-8.92	30
NPT2	Southern	Northern Portugal	2017	41.14	-9.03	47
SPT2	Southern	Southern Portugal	2017	36.84	-8.38	48
NAF	Saharo-Mauritanian	North Africa	2016	20.20	-17.50	57
NES	Western	Northern Spanish shelf	2016	43.59	-2.78	96
MED	Mediterranean	Alboran Sea	2018	36.49	-4.42	49

1144 (*) Stock divisions after Abaunza et al. (2008).

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