Genomic signals of local adaptation in *Eleginops maclovinus* from Northern Chilean Patagonia

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Abstract

Understanding the evolutionary mechanisms that shape the adaptive divergence across

spatially heterogeneous environments is a challenging task for evolutionary ecologists. The

Chilean marine Patagonia is a complex ecosystem with diverse geomorphology and physical-

chemical oceanographic conditions. There is limited research evaluating the interactions

between selective forces and environmental conditions in this area. This study focuses on

identifying the genomic signals of local adaptation of the endemic marine fish, *Eleginops*

maclovinus from Chilean North Patagonia. To achieve this goal, we used an environmental

marine database (temperature, salinity, oxygen, phosphate nitrate and silicate concentration)

with collected from 1995 to 2018 and 11,961 SNPs obtained from 246 individuals from 10

sampling locations across this area. We identified putative adaptive loci using ten

bioinformatic software tools, where five were based on population genetic differentiation

(PGD) and five based on the genotype-environment association (GEA). We identified 392

adaptative loci using PGD and 2,164 associated with at least one of the six environmental

variables analyzed using GEA. A total of 131 loci were shared between the PGD and GEA

approaches, of which 37 were associated with genes involved in the growth, metabolism and

homeostasis. Then, we evaluated the variation of adaptive loci with environmental variables

using polygenic scores and found significant correlations with temperature, salinity, and

oxygen, indicating polygenic selection along environmental gradients. This study highlights

how polygenic selection drives local adaptation in *Eleginops maclovinus* and underscores the

value of integrating genomic and environmental data for conservation in the Patagonian

ecosystem.

Keywords: Patagonian blennie, SNPs, candidate genes, seascape genomics, RAD-seq.

1 Introduction

The main aim of adaptive genomics is understanding the molecular basis of local adaptation in species experiencing spatially heterogeneous environments. Currently, detecting adaptative loci is possible using high-throughput sequencing technologies, which allows thousands of single-nucleotide polymorphisms (SNPs) distributed across the genome (Willette *et al.* 2014), to be genotyped in hundreds of individuals (Manel *et al.* 2016).

Restriction site-associated DNA sequencing (RAD-seq) method harness the high-throughput sequencing technology to genotype individuals (Miller *et al.* 2007; Andrews *et al.* 2016; McKinney *et al.* 2017a). Specifically, this method uses restriction enzymes that cut at specific motifs throughout the genome to reduce their complexity (Altshuler *et al.* 2000; Miller *et al.* 2007; Baird *et al.* 2008). The use of hundreds of thousands of RAD-seq markers for exhibiting the signatures of selection is increasingly applied to marine organisms, offering particular advantages for complex genomes in the absence of a reference sequence (Wang *et al.* 2015; Benestan *et al.* 2016; Cure *et al.* 2017; Dalongeville *et al.* 2018; Hoey & Pinsky 2018; Li *et al.* 2019; Segovia *et al.* 2020; O'Leary *et al.* 2021).

For identifying adaptative loci, there are two predominant genome scan approaches: *(i)* population genetic differentiation (PGD) and *(ii)* genotype-environment association (GEA). The approach based on population genetic differentiation (PGD) identifies loci putatively under selection by primarily comparing the genetic differentiation index (F_{ST}) and other measures (e.g., inbreeding coefficient (F_{IS}) and locus-specific homozygosity excess) for each locus against values predicted by a null model of neutral evolution. If natural selection favors one allele at a particular locus in some populations, F_{ST} at that locus will be large compared

to loci in which among-population differences are the result of neutral demographic

processes (François *et al.* 2016). Contrarily, approach based on genotype-environment

association (GEA) search for associations between environmental variables and allele

frequencies across a set of populations spread over a geographic space (Rellstab *et al.* 2015;

Hoban *et al.* 2016). Alleles involved in local adaptation should occur at higher frequency

where they increase fitness and be at lower frequency where they decrease fitness. Therefore,

it may be possible to identify putative adaptive loci by scanning for alleles that show greater

than average genetic differentiation among populations using GEA approach (Lewontin &

Krakauer 1973; Beaumont 2005). These approaches are based on the premise that it is

possible to distinguish between the genetic signals left by adaptative and neutral processes.

However, algorithms, statistical models, and prior assumptions considerably differ and all of

them are prone to false positive associations (de Villemereuil *et al.* 2014; Rellstab *et al.* 2015;

de Villemereuil & Gaggiotti 2015; François *et al.* 2016). Combining PGD and GEA

outcomes may provide an efficient strategy to detect signatures of local adaptation while

reducing the number of false positives (de Villemereuil *et al.* 2014; Rellstab *et al.* 2015;

Gagnaire *et al.* 2015; François *et al.* 2016).

North Patagonian ecosystems in southern Chile are constituted by extensive and

interconnected system of fjords, channels, gulf, estuaries, and bays that are affected by

physical regimes that may strongly modulate biological productivity (Pantoja *et al.* 2011;

Iriarte *et al.* 2014). It extends approximately across 140,000 km^2 from latitude ~41.5ºS

(Reloncaví Fjord) to latitude ~46.5ºS (San Rafael Lagoon) (Rodrigo 2008).

Oceanographically, Chilean Northern Patagonia is considered a transitional marine system,

where the interaction between marine and fresh waters produces strong vertical and

horizontal gradients in physical-chemical variables (Pickard 1971; Dávila *et al.* 2002; Iriarte *et al.* 2014; Pérez-Santos *et al.* 2014; Schneider *et al.* 2014). These characteristics make Patagonia an excellent natural laboratory to investigate how this seascape heterogeneity results in genomics features. Thus, integrating genomic and environmental data provides a powerful approach for uncovering the processes leading to adaptative divergence in ecologically heterogeneous and connected marine ecosystems. So far, studies focused on seascape genomics in Northern Patagonia are limited (Araneda *et al.* 2016; Canales-Aguirre *et al.* 2016, 2018; Delgado *et al.* 2019; Segovia *et al.* 2020; Canales-Aguirre *et al.* 2022).

Eleginops maclovinus (Cuvier & Valenciennes 1830), is a notothenioid fish (Perciformes) of the monotypic family Eleginopidae (Pequeño 1989). This marine fish is endemic to the coastal temperate and sub-Antarctic waters of South America with a distribution range along the Atlantic and Pacific Patagonian coasts. In the Atlantic Ocean, it is distributed from Uruguay (~35°S) to the Beagle Channel (~54°S), including the Falkland/Malvinas Islands (Gosztonyi & Ae 1974; Eastman 1993), while in the Pacific Ocean, from the Beagle Channel to Valparaíso (~33°S) (Pequeño 1989). This species is a protandrous hermaphrodite with the highest fecundity (~796 eggs g^{-1}) (Brickle *et al.* 2005) among notothenioids (80-300 eggs g^{-1}) (Kock 1992; North 2001). It is euryhaline and can tolerate a broad range of salinity concentrations (5 and 45 PSU), however, extreme salinities can produce dramatic metabolic responses to fuel osmoregulatory tissues (Vargas-Chacoff *et al.* 2014, 2015a). Furthermore, juveniles of *E. maclovinus* have been described having a large eurythermal capacity, with tolerance between 4 and 20°C (Vanella *et al.* 2012; Oyarzún *et al.* 2018; Lattuca *et al.* 2018). All these studies described above have been conducted under controlled conditions, they are focused on one or two environmental variables, and do not

associate with genetic data. The latter leaves a gap on how loci under selection can shape the

adaptative capacity of *E. maclovinus* populations considering the complex environmental

variability in estuaries and fjords.

Studies about the geographic pattern of genetic diversity in *E. maclovinus* are limited

to three (Ceballos *et al.* 2012, 2016; Canales-Aguirre *et al.* 2022). A phylogeographic

research conducted by Ceballos et al. (2012) using mitochondrial DNA sequences for nine

locations did not find any evidence for the structuring of genetic variation into an Atlantic

and a Pacific group. Subsequently, using nine microsatellites it was revealed a low but

significant level of genetic differentiation between Pacific and Atlantic populations and

showed differences between locations in the Pacific coast (Ceballos *et al.* 2016). However,

the authors acknowledge that a single mitochondrial locus or nine microsatellites may not be

robust enough to allow them to recover significant subdivision among geographic regions.

Recently, Canales-Aguirre *et al.* (2022) used 12,050 SNPs in five locations distributed across

Northern Patagonia to identify neutral and adaptive population structure. However, the latter

only included five locations, used a public environmental database Bio-ORACLE

(Tyberghein *et al.* 2012; Assis *et al.* 2018) which is not accurate for fine-scale and fjords

areas like Patagonia, and only three approaches to identify adaptive loci. In addition, its focus

was to identify natural and adaptive loci for conservation and management purposes. To our

knowledge, this is the first study to identify the signal of local adaptation using ten different

bioinformatic approaches and six highly accurate environmental variables in populations of

E. maclovinus along Northern Patagonia.

In this study, we applied RAD-seq sequencing to generate SNPs genotyped in 246

individuals of *E. maclovinus* collected in North Patagonia. Here, we expanded the sampling

locations from Canales-Aguirre et al. (2022), increasing the number to ten locations to better

capture the environmental heterogeneity of their North Patagonian seascape. Our aim was to

detect SNPs putatively under selection using population genetic differentiation (PGD) and

genotype-environment association (GEA) approaches. We tested associations between SNPs

and six environmental variables (temperature, salinity concentration of oxygen, nitrate,

phosphate and silicate) obtained from oceanographic data recorded between 1995 to 2018 in

this area. We implemented a BLASTx search on putative adaptive SNPs identified by both

PGD and GEA approaches to investigate genes with molecular functions potentially involved

in local adaptation. Then, we evaluated the variation of adaptive loci using polygenic scores

where we found significant correlations and supporting polygenic selection in North

Patagonia.

2   Materials and Methods

2.1   Study area and environmental predictors

Chilean Patagonia is one of the most extended fjord regions in the world, located on the southeastern border of South America (Fig.1A), and composed by islands, peninsulas, fjords, and channels (Pantoja *et al.* 2011; Iriarte *et al.* 2014). This region has been affected by ice sheets calving into the ocean, retraction of the sea coast line and decrease in marine water temperatures since the Great Patagonian Glaciation (c. 1 Ma) during the Pleistocene (Rabassa *et al.* 2011). The Northern Patagonia is characterized by interactions between marine and fresh water, generating sharp vertical and horizontal salinity gradients (Dávila *et al.* 2002; Pérez-Santos *et al.* 2014; Schneider *et al.* 2014). The temperature shows a latitudinal gradient, which decreases southward until the Antarctic Polar Front (~55°S) (Iriarte *et al.* 2014). The oxygen concentration in the fjords is the result of horizontal advection of adjacent well-oxygenated Subantarctic Waters (5–6 mL/L) representing the major source of oxygen in the deep layers of the inner seas of Patagonia (Silva & Vargas 2014). In recent years, dissolved oxygen (DO) in the fjords of Chilean Patagonia has decreased by 1–2%, potentially driven by nutrient accumulation and rising temperatures (Linford *et al.* 2023a; b, 2024). The concentrations of inorganic nutrients show a strong seasonal signal, with high nitrate and orthophosphate concentrations during winter, and lower values during spring, presumably caused by a severe increase in primary productivity when light availability in near-surface waters increases (Iriarte *et al.* 2007; González *et al.* 2011; Torres *et al.* 2014; Jacob *et al.* 2014). The concentration of silicic acid shows latitudinal and longitudinal gradient due to surface water mixing between nitrate-rich but Silicic acid-

176   depleted oceanic subantarctic waters and Silicic acid-rich but nitrate depleted continental

177   waters (Aracena *et al.* 2011; González *et al.* 2011; Torres *et al.* 2014).

Figure 1   Geographic area and environmental variables in used.

Map of sampling locations in South America. B) Sampling site locations on North Patagonian fjord ecosystems in southern Chile REL:

Reloncaví Estuary, MAN: Manao, HOR: Hornopirén, CHE: Chepu, CHA: Chaitén, RMB: Raúl Marín Balmaceda, MEL: Melinka Island,

PUY: Puyuhuapi, CIS: Port Cisnes, AYS: Port Aysén: C) Mean surface temperature (ºC) D) Mean surface salinity (g/Kg); E) Mean

surface concentration oxygen (μM); F) Mean surface nitrate (μM); G) Mean surface phosphate(μM); H) Mean surface silicate (μM).

We compiled an environmental database using data collected through the CIMAR Program

of the National Oceanographic Committee of Chile (CONA), from CIMAR 1 (1995) to

CIMAR 24 (2018). The database comprises 1,195 sampling stations located along the inner

sea of Chiloé Island, estuaries, channels, and fjords from the Chilean Patagonia sampled in

spring and winter (http://www.shoa.cl/n_cendhoc/productos/reporte_datos.php) (Fig. S1). At

each sampling station, environmental variables were measured from the surface to the

maximum possible depth. Considering that traditional databases lack high-resolution data

near the coasts (e.g., Bio-Oracle, World Ocean Atlas, Aqua/MODIS), we highlight the use

of information obtained by CIMAR as an effective alternative to detect the environmental

heterogeneity of Northern Patagonia. Our database compiled contains six environmental

predictors: conservative temperature[1] (ºC), absolute salinity[2] (g Kg^{-1}), oxygen (mL L^{-1}),

nitrate (µM), phosphate (µM) and silicate (µM) concentrations. We used Ocean Data View

v.5.4.0. (Schlitzer 2002) for data visualization. We estimated the mean, minimum, maximum

and range of these environmental variables in each of the ten locations for a water column

until 100 m depth using the entire period of the database (1995 – 2018) (Table S1). These

latter, considering that *E. maclovinus* inhabits in coastal and estuarine areas at depths >40 m,

and the spawning occurs in spring season (Brickle *et al.* 2005)

[1] Conservative temperature is a thermodynamic property of seawater. It is derived from the potential enthalpy and is recommended under the TEOS-10 standard (Thermodynamic Equation of Seawater-2010) as a replacement for potential temperature as it more accurately represents the heat content in the ocean.

[2] The Absolute Salinity is the mass fraction of dissolved non-H_2O material in a seawater sample at its temperature and pressure and is expressed in g Kg^{-1}

2.2 Sampling and genotyping

Between November 2018 and April 2019, we sampled 246 individuals of *E. maclovinus* from ten locations throughout the North Patagonian fjord ecosystems in southern Chile (Fig.1B). Five of the ten sampling locations included in this study were previously analyzed by Canales-Aguirre *et al.* (2022). Sampling, DNA extraction, RAD-sequencing library preparation, sequencing with Illumina technology and pipeline of bioinformatic analyses were performed following procedures described by Canales-Aguirre *et al.* (2022).

Loci and samples were filtered iteratively by initially removing poor-quality data, recalculating missing data proportions, and applying increasingly stringent thresholds, following McKinney et al. (2020). The first part of the filtration process consisted in excluding SNPs or individuals based on the following criteria *(i)* minor allele frequencies (MAF) per sampling location, *(ii)* genotype rate per locus, *(iii)* genotype rate per sample. Initially, filtration criterion was MAF ≤ 0.05, genotype rate threshold of 50% per locus and sample. Subsequently, MAF≤ 0.1, genotype rate threshold of 90% per locus and sample. Due to gene duplication events described to have affected the evolution of the notothenioid genome (Xu *et al.* 2008; Chen *et al.* 2008, 2019), *HDplot* (McKinney *et al.* 2017b) was used to identify paralogs, which were removed from further analysis (parameter flags: $H < 0.6$; $|D| < 5$). We removed SNPs in Hardy-Weinberg disequilibrium (p-value < 0.05) if they deviated in three or more locations to avoid potential issues such as inbreeding, population stratification, and genotyping errors in the identified SNPs (Wigginton *et al.* 2005). For reads with more than one SNP, we only kept those with the highest F_{ST}, to reduce the influence from linked loci in the results. The resulting filtered (.vcf) file was converted into the

corresponding file formats for the subsequent analyses by using *PGDSpider* v. 2.1.1.5.

(Lischer & Excoffier 2012).

2.3   Detecting putative loci under selection

We employed ten different software programs to identify putative adaptive loci: five

based on population genetic differentiation (PGD) and five on genotype-environment

association (GEA). We combined multiple approaches to minimize false positive rates and

maximize our chances of detecting genomic signals of local adaptation (Rellstab *et al.* 2015;

François *et al.* 2016). For the PGD approach, we used *fsthet* v.1.0.1 (Flanagan & Jones 2017),

BayeScan v.2.1 (Foll & Gaggiotti 2008), *OutFLANK* v.0.2 (Whitlock & Lotterhos 2015),

PCAdapt v.4.3.3 (Luu *et al.* 2017; Privé *et al.* 2020)*,* and *Arlequin* v.3.5.2.2 (Excoffier &

Lischer 2010) (Table 1). For the GEA approach, we used six environmental variables and

five software programs: *Latent Factors Mixed Model (LFMM)* (Frichot *et al.* 2013),

BayeScEnv (de Villemereuil & Gaggiotti 2015), *Redundancy Analysis (RDA).*(Hair *et al.*

1995; Zuur *et al.* 2010; Capblancq *et al.* 2018; Capblancq & Forester 2021)*, Moran Spectral*

Outlier Detection (MSOD) (Wagner *et al.* 2017), and *Samβada* (Stucki *et al.* 2017; Duruz *et*

al. 2019) (Table 2). In all programs, a 5% False Discovery Rate and respective p-values or

q-values were applied to detect putative adaptive loci. We employed *UpSet* diagrams (Lex

et al. 2014) to visually compare the results from the ten software programs, identifying loci

flagged as outliers by at least two programs as putative adaptive loci. For subsequent

analyses, loci were classified into two categories: putative adaptive loci (under directional

selection) and putative neutral loci, excluding loci under balancing selection.

Table 1 Description and parameters used for software with the PGD approach.

Software	Description	Parameters	References
i) *fsthet*	*fsthet* identifies loci with extreme F_{ST} values by calculating smoothed quantiles for the F_{ST}-heterozygosity distribution, using its own empirical distribution instead of a null model due to changes in population structure.	We used a confidence level of 95% to detect F_{ST} outliers	(Flanagan & Jones 2017)
ii) *BayeScan*	*BayeScan* estimates the probability of selection for each locus using a Bayesian test that distinguishes between locus-specific selection effects (α) and population-specific demography effects (β), by measuring the discordance between global and population-specific allele frequencies.	We ran it with default parameters: 20 pilot runs of 5,000 iterations followed by 50,000 iterations with an additional burn-in of 50,000 and prior odd of 10.	(Foll & Gaggiotti 2008)
iii) *OutFLANK*	OutFLANK calculates a neutral distribution of F_{ST} values and assigns q-values to each locus to detect adaptive loci influenced by selection, without assuming a specific population's demographic history, reducing false positives.	We ran it with default parameters: LeftTrimFraction = 0.05, RightTrimFraction = 0.05, Hmin = 0.1.	(Whitlock & Lotterhos 2015)
iv) *PCAdapt*	*PCAdapt* performs a Principal Component Analysis (PCA) and computes the p-values for each locus to detect putative adaptative loci.	The most appropriate number of clusters K was chosen by testing K=1:10. We detected outliers using K=2 and p-values < 0.05 associated with Mahalanobis distances calculated with the q-value package.	(Dabney *et al.* 2010; Luu *et al.* 2017)
v) *Arlequin*	*Arlequin* calculates observed heterozygosity (H_O) to create a null distribution of F_{ST} values and associated p-values for each locus.	This was executed under a hierarchical island model with 20,000 simulations and 100 demes, which implements FDIST2 methodology.	(Excoffier & Lischer 2010)

246 **Table 2** Description and parameters used for software with the GEA approach**.**

Software	Description	Parameters	References
i) Latent factors mixed model (*LFMM*)	*LFMM* is a hierarchical Bayesian mixed model, which uses latent factors to correct population structure while fitting a linear regression between allele frequencies and environmental variables.	We estimated the number of ancestral populations K before the analysis testing from K 1 to 10 using the LEA v.3.2.0 R package and determined K from the cross-entropy criteria and Cattell's rule from the sNMF output. LFMM analyses were conducted separately for each environmental variable, with 10,000 iterations, and burning of 5,000 and five replicates.	(Cattell 1966; Devlin & Roeder 1999; Frichot *et al.* 2013, 2014; Frichot & François 2015)
ii) *BayeScEnv*	*BayeScEnv* detects putative adaptive loci by identifying loci that show large positive F_{ST} values (outside the neutral model F_{ST} distribution) that are significantly correlated with environmental variables.	We used 20 pilot runs, with 50,000 iterations each and an additional burn-in of 50,000 iterations for each of the six standardized environmental variables.	(de Villemereuil & Gaggiotti 2015).
iii) *Redundancy Analysis (RDA)*	*Redundancy Analysis (RDA) is a* multivariate ordination method to detect loci putatively under selection based on correlations with environmental variables.	We calculated the variance inflation factor (vif.cca function) to exclude variables with a value ≥ 5, and conducted linear regressions between allele frequencies and environmental variables at each locus. Using PCA, we produced ordination axes, identified putative adaptive loci (±3 SD from the mean), and evaluated the correlation using the Pearson correlation coefficient.	(Hair *et al.* 1995; Zuur *et al.* 2010; Rellstab *et al.* 2015; Forester *et al.* 2018)
iv) *Moran spectral outlier*	*MSOD* uses Moran eigenvector maps (MEM) to quantify the distribution of allele frequencies	We compared the power spectrum of each locus to the average power spectrum to identify outliers using z-scores and a cut-off of 0.05. Subsequently, we applied Moran	(Dray *et al.* 2006, 2021; Wagner & Dray 2015;

	detection (MSOD)	across a range of spatial scales represented by MEM spatial eigenvectors	spectral randomization (MSR) with 999 permutations using adespatial.	Wagner *et al.* 2017)
v)	*Samβada*	*Samβada* is a spatial approach that uses multiples univariate logistic regression models to identify locus-environment associations and at the same time measures spatial autocorrelation	We coded individuals for the presence or absence of each SNP allele and measured the association with environmental parameters across sites using multiple logistic regression. P-values from these regressions were calculated via a Wald Score Test and compared to a χ^2 distribution with one degree of freedom.	(Wald 1943; Stucki *et al.* 2017; Duruz *et al.* 2019)

2.4　Genetic diversity and population structure

For the adaptative and neutral datasets, we determined, described and compared overall

and population-specific genetic diversity. The expected heterozygosity (H_E), observed

heterozygosity (H_O) and inbreeding coefficient (F_{IS}) were calculated for each locus and

locations using *Hierfstat* v 0.04.10 R package (Goudet 2005). We assessed the genetic

differentiation with pairwise F_{ST} (Weir & Cockerham 1984) comparisons between locations

using the R package *StAMPP* v.1.6.2. (Pembleton *et al.* 2013) and the statistical significances

were determined using 10,000 permutations.

We used two approaches to investigate the genetic structure: Bayesian clustering

program *Structure* v.2.3.4. (Pritchard *et al.* 2000) and discriminant analysis of principal

components (DAPC) (Jombart *et al.* 2010) using r package *adegenet* v.2.1.5. (Jombart 2008;

Jombart & Ahmed 2011). *Structure* is a Bayesian method, which uses a Markov chain Monte

Carlo (MCMC) algorithm based on the Gibbs sampler algorithm to identify the number of

putative genetic clusters. This method assumes Hardy-Weinberg equilibrium (HWE) and

linkage equilibrium among loci in sample population individuals (Pritchard *et al.* 2000). We

conducted the analysis using ten replicates per K value for K=1 to K=10, with 10,000 burn-

in steps and 100,000 MCMC-sampling steps. The runs were collected, sorted and merged

and functions in the R package *pophelper* (Francis 2017) to produce assignment barplots.

The most likely number of genetic clusters (K) was chosen using the ΔK method (Evanno *et*

al. 2005). The DAPC method identifies and describes clusters of genetically related

individuals from genomic datasets and allows the optimal visualization of between-

population differentiation in multivariate space (Jombart *et al.* 2010). First, we carried out a

principal component analysis (PCA), and used the principal components (PCs) thus produced

as synthetic variables for a discriminant analysis, as outlined in Jombart et al. (2010). The

optimal number of principle components was determined by cross validation (*xvalDapc*

function (Jombart *et al.* 2010)) to avoid overfitting.

We also conducted isolation by distance (IBD) analyses to test whether genetic distance

increased with geographic distance with adaptative and neutral data. We assessed the

correlation between the linearized pairwise F_{ST} matrix and the Euclidean (geographic)

distance matrix using a Mantel-test as implemented in the R-package *vegan* (Oksanen *et al.*

2016). Statistical significance was assessed with 999 permutations of the data. To test

whether the correlation between genetic and geographic distances is a result of a continuous

or distant patchy cline of genetic differentiation, local densities of distances to disentangle

both processes were plotted using 2-dimensional kernel density estimation (*kde2d* function)

of *MASS* R-package (Venables & Ripley 2013).

2.5   Additive polygenic scores (APS)

We assessed cumulative adaptive genetic variation by calculating additive polygenic

scores (APS) associated with each environmental variable (Gagnaire & Gaggiotti 2016;

Babin *et al.* 2017). For this analysis, we prioritized three biologically important

environmental variables: annual maximum temperature (°C), annual maximum salinity

(g/Kg), and annual minimum oxygen concentration (µM) (Holliday 1969; Nikinmaa & Rees

2005; Brian *et al.* 2008; Vanella *et al.* 2012; Vargas-Chacoff *et al.* 2014). We prioritized the

maximum values of temperature and salinity considering the effect that climate change has

on them. According to the Community Climate System Model 3.0 (CCSM3) of the National

Center for Atmospheric Research (NCAR), a 2.5°C increase in the temperature of Chilean

coastal areas is predicted by 2055 (Silva *et al.* 2016). Similarly, the Center for Oceanographic

Research in the Eastern South Pacific (COPAS) detected an increase of 0.25 in salinity in the

upper layers of the central-southern coast of Chile (Schneider *et al.* 2017)We also considered

the minimum oxygen concentration, as hypoxic conditions negatively impact fish fitness

(Richards 2009; Wu 2009). All these variables have an impact of the adaptive genetic

diversity of several species (Pauls *et al.* 2013; Bernatchez 2016)..

We assessed the relationship between each of the 131 putative adaptative loci and each

environmental variable. We used genotypes (0,1,2) to determine the score for each favored

allele; if the slope is positive, and (2,1,0) if negative (Babin *et al.* 2017). We calculated APS

by summing the scores of favored alleles associated with each environmental variable.

Finally, we evaluated the relationship between individual APS and environmental variables

using three models: linear, quadratic, and null, determining the best fit based on the lowest

Akaike Information Criterion (AIC) value.

2.6   Loci annotation and Candidate Gene Identification

Loci identified as putative adaptative loci (see results section 3.5) were functionally

annotated through Blast2GO v.6.0.3 (Conesa & Götz 2008). We compared the sequences

against bony fishes (taxa: 7898, Actinopterygii) to determine whether any putative adaptative

loci were associated with genes that may be related to environmental tolerances (e.g., heat

shock, immune system proteins, osmoregulation). We applied the BLASTx algorithm with a

cut-off of E-value $\leq 1\mathrm{x}10^{-5}$, homology of sequences of more than 75% and other parameters

were used as default. Subsequent analysis only considered sequences with functional

annotation (blasted, mapped and annotated).

3   Results

3.1   RAD sequencing and data filtering

After quality control, 2,359,848 (511,190 RADtag) putative loci were obtained for Stacks pipeline. Using the iterative filtering process (MAF and missing data), we identified a total of 2,328,560 (98.67%) putative singletons and 386 putative paralog loci; the latter were removed (1.24%). Further filtering by HWE discarded 13,215 loci (42.96%) and one SNP per tag (the highest F_{ST}) discarded 5,582 SNPs (31.82%). A total of 11,961 SNPs (5.01%) were retained, from 202 individuals across the 10 sampling locations (Table 3).

Table 3   Description of steps used to filter SNPs

Filtering steps	Sample size	Number of loci
SNP after STACKS	246	2,359,848
[1]MAF >= 0.05	246	1,494,046
[1]Genotyped by locus (50%)	246	167,752
[1]Genotyped by sample (50%)	236	167,752
[1]Genotyped by locus (90%)	236	54,694
[1]Genotyped by sample (90%)	202	54,694
[1]MAF >= 0.1	202	31,144
Singletons (HDPlot H=0.6; D=5)	202	30,758
Hardy Weinberg equilibrium in 100% locations	202	17,543
one SNP by tag	202	11,961

[1] Steps included in the iterative filtering process

3.2   Identification of loci putatively under selection

From population genetic differentiation (PGD) and genotype-environment association

(GEA) we identified different putative adaptative loci. Out of the 11,961 loci, we identified

392 loci under positive selection by PGD, and 2,164 loci using the GEA approach. A total of

131 loci were identified as shared between the PGD and GEA approaches (Figure 2).

Figure 2   UpSet diagram: matrix layout for all intersections of software used. A) Loci

detected by software based on Population Genetic Differentiation (PGD), B) Loci detected

by software based on Environment Genotype-Association and C) Loci shared between PGD

and GEA. The set size (horizontal bars) indicates the total number of loci detected by that

software. The intersection size (vertical bars) indicates the number of loci per set intersection,

the purple dots represent the set of unique loci detected only by that software and no other

one, while the purple dots connected by a line represent loci shared by two, three, or more

software. The *OutFLANK, BayeScan, BayeScEnv,* and *Samβada* software are excluded since

they did not identify any adaptative loci**.**

Within PGD approaches, *fsthet* detected 334 SNPs (2.79%) under balancing selection and

313 SNPs (2.62%) under divergent selection (Figure S2). *Arlequin* identified 270 SNPs under

divergent selection (2.26%) (Figure S3). *PCAdapt* detected 15 SNPs (0.13%) under divergent

selection (Figure S4). *BayeScan* and *OutFLANK* found no outlier. For GEA approaches,

LFMM identified 1,785 loci under divergent selection, with the highest number of

associations to nitrate concentration (n = 684 loci) (Figure S5). *MSOD* detected 377 loci,

with the highest number of associations to temperature (n = 377 loci) (Figure S6). Finally,

RDA detected 83 loci under divergent selection, with the highest number of associations to

silicate (33 loci) (Figure S7). *Samβada* and *BayeScEnv* did not detect any correlation with

environmental variables. Overall, across GEA approaches, the nitrate concentration (n =

844), temperature (n = 748), and salinity (n = 703) were the environmental variables with

which most of the loci were correlated (Figure 3). Of the total of 2,164 that correlated with

each of the six environmental variables, 783 of the loci were significantly associated with

more than one environmental variable.

Figure 3   UpSet diagram: matrix layout for all intersections of environmental variables. The set size (horizontal bars) indicates the total number of loci correlated with that environmental variable. The intersection size (vertical bars) indicates the number of loci per set intersection, the purple dots represent the set of unique loci correlated only with that

environmental variable and no other one, while the purple dots connected by a line represent

loci shared by two, three, or more variables.

3.3   Assessment of genetic diversity and population structure

Summary statistics of genetic diversity revealed similar values for each location within

each dataset (Figure S8). Most locations showed lower values of observed heterozygosity

(H_O) than expected heterozygosity (H_E), and positive values of inbreeding coefficient F_{IS}. For

the adaptative loci dataset, H_O 0.3364 ± 0.0078 (C.I.$_{95\%}$), H_E $0.3551 \pm 0,0068$(C.I.$_{95\%}$), and

F_{IS} 0.0528 ± 0.0120 (C.I.$_{95\%}$). For the neutral dataset, H_O 0.3414 ± 0.0074 (C.I.$_{95\%}$), H_E 0.3485

$\pm 0,0063$ (C.I.$_{95\%}$), and F_{IS} 0.0206 ± 0.0117 (C.I.$_{95\%}$).

Pairwise-F_{ST} comparisons showed higher levels of genetic differentiation for adaptive

loci (global F_{ST} = 0.0419 ± 0.0042 CI$_{95\%}$), than neutral loci (global F_{ST} = 0.0004 ± 0.0004

CI$_{95\%}$). For adaptative loci, all pairwise differentiation values were highly significant (p-value

≤ 0.0004). While in neutral loci, 26 of 45 pairwise differentiation values showed significant

p-values (Table 4).

Table 4 Pairwise F_{ST} (below diagonal) and p-value (above diagonal).
Values between sampling site locations from North Patagonian fjord ecosystems in southern
Chile using A) adaptative loci and B) neutral loci. Statistical significance was considered
below 0.05 p-value.
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A)	REL	MAN	HOR	CHE	CHA	RMB	MEL	PUY	CIS	AYS
REL	-	0	0	0	0	0	0	0	0	0
MAN	0.0672	-	0.0004	0	0	0	0	0	0	0
HOR	0.0495	0.0126	-	0	0	0	0	0	0.0004	0
CHE	0.0653	0.0443	0.0493	-	0	0	0	0	0	0
CHA	0.0665	0.0502	0.0503	0.0531	-	0	0	0	0	0
RMB	0.044	0.0302	0.0227	0.0502	0.0502	-	0	0	0	0
MEL	0.0755	0.0343	0.0296	0.0521	0.0539	0.0313	-	0	0	0
PUY	0.0467	0.0403	0.0266	0.0546	0.0408	0.0286	0.033	-	0	0
CIS	0.0592	0.0338	0.0210	0.0546	0.0435	0.0248	0.0355	0.0234	-	0
AYS	0.0472	0.0442	0.0199	0.0553	0.0456	0.0269	0.0432	0.0263	0.0292	-

B)	REL	MAN	HOR	CHE	CHA	RMB	MEL	PUY	CIS	AYS
REL	-	0	0	0.9997	0.7841	0.1541	0	0.0095	0.0763	0.0121
MAN	0.0018	-	0.0004	0.9909	0.3254	0	0	0	0.0008	0
HOR	0.0017	0.0011	-	0.998	0.9681	0.0018	0.0023	0	0.0174	0
CHE	-0.0018	-0.0011	-0.0013	-	1	1	0.9814	0.9893	0.9994	0.9887
CHA	-0.0394	0.0002	-0.0789	-0.0036	-	0.8519	0.8760	0.2556	0.9314	0.4809
RMB	0.0004	0.0016	0.0011	-0.0022	-0.0466	-	0.0003	0.0489	0.3379	0
MEL	0.0019	0.0019	0.0011	-0.0011	-0.524	0.0013	-	0	0.0070	0
PUY	0.0010	0.0020	0.0019	-0.0011	0.0297	0.0006	0.0020	-	0.0004	0.0002
CIS	0.0006	0.0012	0.0007	-0.0017	-0.0684	0.0002	0.0010	0.0014	-	0.0006
AYS	0.0009	0.0017	0.0015	-0.0010	0.2420	0.0012	0.0016	0.0012	0.0012	-

Evanno's methods showed that the likely number of K clusters was K = 2 for adaptative

loci (Figure S9) and K = 7 for neutral loci (Figure S10). For adaptative loci, a visual

inspection revealed a clear separation of individuals for REL, MAN and MEL using K=2.

We observed an additional cluster which appeared to correspond to individuals from CHE

and CHA using K = 4 (Figure 4A). For neutral loci, *Structure* showed no clear geographical

genetic structure (Figure 4B). Results from DAPC analyses revealed similar patterns to

Structure (Figure 5). The adaptative dataset showed that the REL, CHA and CHE were

separate from all other locations. The first two axis explain 44.9% of the total variation. For

the neutral dataset, PET, CHA and CHE overlapped greatly with the other locations and much

less structure overall was present, although MAN seemed to be less like the other nine

locations. The first two axis explain 40.5% of the total variation (Figure 5).

Figure 4   Results from clustering analysis of the *E. maclovinus.*

Based on A) 131 putatively adaptative loci and B) 9,536 putatively neutral loci. Each vertical

line represents an individual. The colors represent the proportion of inferred ancestry from

the K ancestral populations. Black lines separate the individuals from different locations. ΔK

calculated for K ranging from 1 to 10, with each K repeated tenfold.

Figure 5 Scatterplot of discriminant analysis of principal component (DPCA).

A) adaptative loci and B) neutral loci. Axes correspond to the first two discriminant functions

and the circles represent individuals colored-coded by sampling location.

The isolation by distance (IBD) analysis for adaptative loci revealed a moderate

significant isolation by distance pattern (Mantel test: $r = 0.3458$, p-value $= 0.0320$, Figure

6A), where the scatterplot of local densities of distances showed only one consistent cloud

of points, indicating a continuous cline of genetic differentiation. For neutral loci, we found

no detectable relationship between geographic distance and genetic distance (Mantel test: r

= -0.0435, p-value = 0.5760, Figure 6B).

Figure 6 Isolation-by-distance IBD among locations. A) 131 adaptative loci and B)
9,536 neutral loci. Local density of points plotted using two-dimensional kernel density
estimation. The black line is the trend estimated by Loess regression red line indicates the
linear regression. Color gradient represents density of observations (low = blue, red = high).

3.4   Additive polygenic scores (APS)

The correlations between additive polygenic scores (APS) and environmental variables

were all significant ($p < 0.001$). The R^2 values ranged from 0.29 for maximum temperature

to 0.26 for both maximum salinity and minimum oxygen concentration. For all

environmental variables, APS decreased with increasing temperature, salinity, and oxygen

levels, respectively (Figure 7). The correlation between APS and environmental variables

was best represented by linear models, except for maximum temperature, which fitted better

with a quadratic model (Figure 7). The significance and R^2 values of the correlations

calculated for mean and minimum temperature, salinity, and oxygen concentration yielded

similar results, albeit with slight differences among subsets (Figure S11-S13).

Figure 7   Correlations between additive polygenic scores (APS) based on A) Annual

maximum temperature B) Annual maximum salinity and C) Annual minimum oxygen and

131 putative adaptative loci. Correlation coefficient (R^2) and p-values and AIC are presented

for each variable.

3.5   Gene ontology of candidate genes

The BLASTx analysis of the contigs containing the 131 adaptative loci shared by PGD

and GEA against bony fish genomes resulted in significant hits (E-value $\leq 1 \times 10^{-5}$) for 790

fish species. The BLASTx similarity results showed that 51 of the 131 contigs containing

adaptative loci corresponded to known proteins in the non-redundant (NR) database (E-value

$1e^{-5}$), of which 36 contigs were functionally annotated. The functional categorization of the

annotated sequences involved biological process, cellular components, and molecular

function (Figure 8).

Figure 8   Putative functional categorization and distribution according to Blast2GO.

A second BLASTx analysis with all 2,425 putative loci detected 512 contigs that were

functionally annotated. Of these 512 sequences, three were already known to potentially play

a role in the local adaptation (Table S1). The sequences of the *SNP 6291_214* and *44791_149*

and belong to the solute carrier (SLC) family. SLC genes encode transmembrane transporters

for inorganic ions, amino acids, neurotransmitters, sugars, purines and fatty acids, and other

solute substrates (Dorwart *et al.* 2008). The SNP 42_59 is located in the vtgC (13489) gene,

which is a member of the vitellogenins (Vtg) family. In egg-laying vertebrates such as fish,

estrogens activate the hepatic synthesis of vitellogenin (Vtg) (Sumpter & Jobling 1995; Brian

et al. 2008).

4 Discussion

This study provides comprehensive evidence for genomic signatures of local adaptation, of *Eleginops maclovinus* to heterogeneous environmental conditions in Northern Patagonia. We conducted a seascape genomics study using 11,961 RAD-sequencing markers and detected 2,164 putative adaptative loci through population genetic differentiation (PGD) and genotype-environment association (GEA) approaches. Using 131 loci shared by PGD and GEA, we observed that the putative adaptive loci collectively exhibit a highly significant association with various environmental variables, indicating polygenic selection. Additionally, we found contrasting patterns of genetic variation in neutral and adaptive loci that reinforce the possible local adaptation among high levels of gene flow. Subsequently, we detected one gene related to thermal adaptation and two genes that play a role in the osmoregulation. In the following, we describe how our study contributes to understanding the role of environmental heterogeneity as a driver of spatial patterns of putative adaptive genetic variation that could influence the local adaptation in *E. maclovinus*. Furthermore, we discuss methodological issues about software for detecting putative adaptive loci. Finally, we propose groups based on adaptative divergence for creating management units.

4.1 Evidence for spatially divergent selection

Evidence suggests thermal adaptation *E. maclovinus*, which juveniles exhibit significant eurythermal capacity, tolerating temperatures between 4 and 20°C (Vanella *et al.* 2012; Oyarzún *et al.* 2018, 2021). Short-term exposure to 4°C may reduce fitness (Vanella *et al.* 2012), while long-term exposure to 15-20°C elevates metabolism with tissue-specific

modulation, without adverse physiological effects. Conversely, temperatures exceeding 25°C

detrimentally impact physiology, growth, food intake, and energy substrates, mainly

triglycerides (Oyarzún *et al.* 2021). These studies focus on juveniles; thus, temperature

effects on early-stage adaptation remain uncertain for this species. Considering *E. maclovinus*

spawns in spring (Brickle *et al.* 2005), higher temperatures could expedite maturation (e.g.

Scott & Pankhurst 1992; Shimizu 2003; Hokanson 2011). Temperature affects reproductive

processes' thermal inhibition (Gillet *et al.* 1996; Tveiten *et al.* 2001; Okuzawa *et al.* 2003;

Soria *et al.* 2008; Pankhurst & King 2010), including egg mortality (Van Der Kraak &

Pankhurst 1997; Gagliano *et al.* 2007; Janhunen *et al.* 2010), embryonic development rate

(Pauly & Pullin 1988; Rombough 1997; Smith *et al.* 2015), and larval fish metabolism,

growth, and development (Houde 1989; Blaxter 1992; Benoît *et al.* 2000). Future research

should examine temperature-induced shifts in selective mortality in early stages of *E.*

maclovinus. Additionally, temperature may influence sexual inversion in protandrous

hermaphrodites, as proposed by Baroiller et al. (1999). For instance, elevated temperatures

increased gonadal aromatase levels, promoting female proportion in *Lates calcarifer,* another

protandrous hermaphrodite (Athauda *et al.* 2012). This latter, although not identified in *E.*

maclovinus, can be taken as a research question for future studies. The detection of

potentially adaptive loci associated with temperature supports molecular adaptation via gene-

related heterogeneity and gene duplication events, likely preserved in response to thermal

fluctuations. Unlike most Antarctic notothenioids, *E. maclovinus* blood shows high

hemoglobin (Hb) multiplicity, with one cathodal (Hb C) and two anodal components (Hb 1,

Hb 2) (Coppola *et al.* 2010). This Hb diversity may facilitate oxygen transport adaptation to

environmental changes and metabolic demands (Weber 1990; di Prisco & Tamburrini 1992;

Feuerlein & Weber 1994; Weber *et al.* 2000; Fago *et al.* 2001). The maintenance of multiple

Hbs in response to temperature differences in temperate waters, significantly larger than in

Antarctica, is likely (Coppola *et al.* 2010; di Prisco *et al.* 2012). Moreover, the absence of

antifreeze-glycoprotein (AFGP) coding sequence in *E. maclovinus* indicates no adaptation to

near-freezing temperatures (Cheng *et al.* 2003).

Evidence suggests salinity acts as a selective agent driving genetic differentiation in *E.*

maclovinus populations. Canales-Aguirre et al. (2022) identified associations between

adaptive loci and environmental variables (salinity, oxygen, primary production, current

velocity) in *E. maclovinus*. Juveniles are considered euryhalines, adapting to salinities

between 5 and 45 PSU (Vargas-Chacoff *et al.* 2014, 2015a). An inverse relationship between

the trypsin chymotrypsin−1 ratio (T/C ratio) and salinity was observed (Vargas-Chacoff *et*

al. 2015b). The T/C ratio, linked to growth and nutritional status (Rungruangsak-Torrissen

et al. 2006; Rungruangsak-Torrissen 2007), suggests better food utilization in low salinity,

enhancing physiological energy. It is possible that food utilization and efficiency is better in

E. maclovinus acclimated to low salinity, leading not to better growth, but to increased energy

available for physiological processes such. Osmoregulation in low salinity consumes energy,

affecting growth (Vargas-Chacoff *et al.* 2015b). Euryhaline fish adapt to varying osmotic

challenges (Laverty & Skadhauge 2012). Climate change, habitat degradation, and

anthropogenic activities increase salinity stress (Kültz 2015). Climatic models for Northern

Patagonia predict reduced precipitation and increased temperature, affecting hydrological

regimes and salinity (Aguayo *et al.* 2019). The salinity variability in microenvironments can

have multiple impacts on reproduction, including DNA damage from salt stress or oxidative

stress (Dowling & Simmons 2009), sperm motility (Beirão *et al.* 2018), incomplete egg

binding (Herberg *et al.* 2018), disrupted egg activation (Ginsburg 1963) and egg penetration

issues (Yanagimachi *et al.* 2017), and cause problems during DNA recombination (O, Wai-

sum *et al.* 2006).

Our results indicate oxygen concentration exerts selective pressure on *E. maclovinus.*

Environmental oxygen variability has been crucial in fish evolution, influencing anatomical,

behavioral, and physiological adaptations for oxygen acquisition and delivery (Nikinmaa &

Rees 2005). Fish have evolved to balance oxygen delivery while avoiding toxic reactive

oxygen species (Taylor & McElwain 2010; Dong *et al.* 2011). They exhibit diverse

hemoglobin structures and functions for oxygen transport (Powers 1974; Jensen & Weber

1985; Weber 1990; Jensen *et al.* 1998). Hemoglobin diversity in fish, including E.

maclovinus, likely results from selection pressures like low oxygen levels(Nikinmaa & Rees

2005). Structural and functional analyses of *E. maclovinus* hemoglobin (Hb1Em) showed

higher O_2 affinity compared to *T. bernacchii,* indicating differences in temperature sensitivity

and O_2 release between sub-Antarctic and Antarctic notothenioids (Coppola *et al.* 2012).

Silva & Vargas (2014) identified hypoxic zones in Northern Patagonia's fjords (1.1–2.0 mL

L-1); we emphasize the importance of studying the ability of to adapt to hypoxic conditions

in *E. maclovinus* until now unclear.

In this study, we identified putative adaptive loci associated with the concentration of

nitrate, phosphate, and silicate. Although there is no direct relationship between nutrient

concentration and adaptive capacity in *E. maclovinus,* the nutrients could be a proxy for

primary productivity. The productivity cycle of Patagonian channels and fjords has been

divided into two seasons: a productive season characterized by a marked spring diatom bloom

followed by pulsed productivity events dominated by a sequence of different phytoplankton

species, and then a non-productive season that predominantly features small phytoplankton

cells (Iriarte *et al.* 2007; Czypionka *et al.* 2011; Montero *et al.* 2011; Paredes & Montecino

2011). During the productive season and in the absence of persistent hydrodynamic

phenomena that force nutrient replenishment in the euphotic zone, the supply of nitrate and

phosphate to the well-lit surface layer would be the chief limiting factor for phytoplankton

growth (Pizarro *et al.* 2000; Iriarte *et al.* 2007). The increased freshwater discharge into the

Puyuhuapi channel during non-productive season coincides with increases in the

concentrations of silicic acid and dissolved organic matter concentrations, as well as an

increase in underwater light attenuation (Montero *et al.* 2017). A GEA analysis of Canales-

Aguirre (2022) provided evidence for local adaptation associated to primary production on

E. maclovinus likely connected to food availability (Gove *et al.* 2016; Fox *et al.* 2018).

Similarly, multiple studies detected putative adaptive loci associated with chlorophyll

(indicator of primary productivity), for example Boulanger *et al.* (2022) with *Mullus*

surmuletus, Cayuela *et al*. (2020) with *Mallotus villosus,* Maselko et al. (2020) with *Sebastes*

alutus and Diopere *et al.* (2017) with *Solea solea*.

4.2   **Additive Polygenic Scores (APS)**

For the Additive Polygenic Scores (APS) (Gagnaire & Gaggiotti 2016; Babin *et al.*

2017), we observed a strong correlation between polygenic scores and maximum temperature

and salinity, and minimum oxygen. This suggests that *E. maclovinus* populations might be

locally adapted to these environmental variables through a polygenic architecture. As we

highlighted in the previous section (See 4.1.), we detailed the impact of these environmental

variables on fish survival and development.

We found negative correlations between individual APS and all environmental

variables, which could indicate directional selection, where different alleles are maintained

in distinct environments with a relatively constant selection force along the environmental

gradient. Specifically, the relationship between APS and maximum temperature was best

represented by a quadratic model, suggesting stabilizing selection at intermediate

temperatures. However, caution is advised with this conclusion as the difference in AIC

between the linear and quadratic models was minimal (AIC $_{linear}$ = 1577.8, AIC $_{quadratic}$ =

1577.7).

The impact of temperature on additive genetic variation agreed with observations in

the adaptive population structure membership plots, where the REL population in the

northernmost latitudinal range with warmer temperatures separates from the other nine

populations. For salinity and oxygen concentration, the linear model better explained the

relationship with APS, indicating relatively constant selection along the salinity and oxygen

gradients. These results are consistent with the observed adaptive population structure, where

the CHE and CHA populations differentiate from the rest, with the CHE population exposed

to higher oceanic salinities. Lastly, the minimum oxygen concentration is critical for fish

reproduction and development, as hypoxia can delay embryonic development and hatching

(Richards 2009; Wu 2009). However, it remains unknown how hypoxia affects the genetic

expression and reproductive development of *E. maclovinus*.

While the APS approach has potential for evaluating the cumulative effect of

potentially adaptive loci in complex environments, it has limitations. Without a reference

genome, we cannot rule out the possibility that some potentially adaptive loci are linked. An

annotated *E. maclovinus* genome would allow us to identify matches between adaptive loci

and selected genomic regions, providing insights into specific genes involved in local

adaptation (Manel *et al.* 2016). Additionally, comparing genotypic variation in potentially

adaptive loci with individual phenotypic data is necessary to establish genotype, phenotype,

and fitness associations (Barrett & Hoekstra 2011).

Our results suggest that maximum temperature, maximum salinity, and minimum

oxygen concentration may drive spatially variable selection in *E. maclovinus* populations.

Temperature, salinity, and oxygen concentration impact the environmental gradient

experienced by eggs and larvae (Holliday 1969; Blaxter 1992; Brickle *et al.* 2005; Wu 2009;

Lehtonen & Kvarnemo 2015), potentially influencing local adaptation. Future research

should aim to detect molecular signatures of selection within a multilocus quantitative

genetics framework to delve deeper into the genetic architecture of polygenic adaptive traits.

4.3   Finding functional genomics of candidate genes for local adaptation

Using BLASTx to identify genes linked to adaptation through a gene ontology

approach (Haasl & Payseur 2016; Manel *et al.* 2016), we found a total of 2,425 potentially

adaptative loci with 131 loci shared by PGD and GEA, only 512 and 36 loci were successfully

annotated, respectively. In the 131 adaptative loci, only 39.69% of loci matched known

sequences in the public database, highlighting the limitation of RAD sequencing in

identifying candidate genes without a reference genome (Davey *et al.* 2013; Andrews *et al.*

2016; Catchen *et al.* 2017; McKinney *et al.* 2017a). Both adaptive datasets revealed loci

linked to life-history traits, including thermal adaptation and osmoregulation.

Focusing on life history traits, we identified five key genes: *opa1, Stx16, FMVIA, foxj1b,*

and *Lamc3*. The gene *opa1* regulates mitochondrial fusion and fission, preventing Cyc-c

leakage and inducing apoptosis, which is linked to kidney injury (Song *et al.* 2017; Yingjie

et al. 2019; Herkenne *et al.* 2020; Wang *et al.* 2020; Cui *et al.* 2021; Li *et al.* 2022). The

Stx16 is involved in the development of inducible defense structures in *Daphnia pulex* under

predation stress (Spanier *et al.* 2010). High *FMVIA* expression in the retina of striped bass

suggests a role in retinal motility (Breckler *et al.* 2000). The *foxj1b* gene is essential for

normal laterality in zebrafish embryos, affecting heart and organ positioning (Tian *et al.*

2009). Finally, *Lamc3* plays a role in the development of the parachordal chain in zebrafish

(Eve & Smith 2017).

Expanding our search to adaptive loci, we identified three key genes: *vtg3* (linked to

temperature adaptation) and *slc39a6, slc8a2a* (associated with salinity adaptation). *Vtgs* are

specialized large lipid transfer proteins produced by the liver and transported to the ovary

(Babin *et al.* 2007). The *Vtg* family includes type-I (*Vtg 1, 4, 5, 6, and 7*), type-II (*vtg2* and

vtg8), and type-III (*vtg3*) (Yilmaz *et al.* 2018). Despite being a minor and divergent form,

vtg3 significantly contributes to zygote and early embryo development (Yilmaz *et al.* 2019).

Increased temperature elevates estrogen-induced *vtg* expression in juvenile *Salmo trutta*

(Körner *et al.* 2008), suggesting a similar role for *vtg3* in *E. maclovinus* related to thermal

adaptation. The gene *slc39a6 (LIV-1)* maintains zinc homeostasis, essential for cellular

growth and differentiation (Taylor & Nicholson 2003). Low salinity increases *slc39a6*

expression in tiger puffer (*Takifugu rubripes*) (Jiang *et al.* 2020), and intracellular zinc

homeostasis controls T-cell survival in zebrafish by inhibiting apoptosis (Zhao *et al.* 2019).

The *SLC8* gene family (*NCX*) maintains cellular Ca2+ homeostasis (Khananshvili 2013).

Slc8a2a, an isoform identified in zebrafish, is key for Ca2+ excretion by seawater fish

kidneys, with upregulated transcription during seawater acclimation (Liao *et al.* 2007; Islam

et al. 2011). These genes indicate positive selection associated with salinity in *E. maclovinus*

in Northern Patagonia.

We did not identify genes whose functions are supposed to be affected by the

concentration of oxygen or nutrients (nitrate, phosphate and silicate). However, many outlier

sequences did not have BLASTx results and may represent important proteins that have not

been adequately annotated in fish to date.

4.4   Comparing approaches for detecting genomic signal of selection

The outcomes from the ten programs varied widely in terms of the number and identity

of adaptative loci detected in this study. Such variations among methods were expected since

their algorithms and prior assumptions considerably differ. Combining population genetic

differentiation (PGD) and genotype-environment association (GEA) approaches to detect

adaptative loci of local adaptation not only reduces false-positive discoveries, but also

maximizes our chances of detecting potential signals of selection (François *et al.* 2016).

Overall, we found that GEA approach identified more adaptative loci (2,164 loci) than PGD

approach (392 loci). These outcomes are in agreement with a simulation study demonstrating

that the GEA approach has more power to detect loci under divergent selection than PGD

approach (de Villemereuil *et al.* 2014), which is not surprising given that GEA uses multiple

environmental variables (six variables in our case) to depict signals of selection. This trend

has been reported in multiple studies focused on detecting signs of local adaptation in marine

populations (Benestan *et al.* 2016; Flanagan *et al.* 2016; Dalongeville *et al.* 2018; Bekkevold

et al. 2020; Cayuela *et al.* 2020; Segovia *et al.* 2020).

Inside of PGD approach, *fsthet* detected the largest number on putative adaptative loci

(n= 313 loci). This software generates smoothed quantiles for the F_{ST}–heterozygosity

relationship from the empirical distribution (Flanagan & Jones 2017). Using an empirical

distribution itself has the advantage of not assuming a particular distribution or model of

evolution. For example, *FDIST2* and *LOSITAN* used an idealized island model, which would

not be appropriate considering the geography of Northern Patagonia. Besides, this feature is

appealing because the F_{ST}–heterozygosity distribution changes based on many potentially

unknowable features of population structure (Flanagan & Jones 2017). Though, *fsthet*

decreased the average proportion of false positives compared with *LOSITAN*, we did not find

any studies that compared the effectiveness of *fsthet* with respect to other software that

considers the hierarchical population structure, such as *Arlequin*. The *fsthet* program is not

recommended when migration rates are low because the neutral loci saturate the full

distribution of potential F_{ST}–heterozygosity values (Flanagan & Jones 2017). In our case, the

relative migration rates showed high gene flow between locations of *E. maclovinus* (Canales-

Aguirre *et al.* 2022). The *Arlequin* identified 270 loci (200 shared with *fsthet*). However,

results of a simulation study showing that *Arlequin* consistently found more outliers and had

the highest type I and type II errors in their simulation scenarios in compared with other

methods such as *BayeScan*. Simulated data suggested that *Arlequin* does not perform well in

situations where adaptive variation contrasts with patterns of neutral variation, likely due to

the effect of hierarchical grouping based on neutral markers (Narum & Hess 2011). The

PCAdapt was the more conservative (n= 15 loci), whereas *BayeScan* and *OutFLANK* did not

detect any outliers. Population structure and the presence of admixture individuals affect the

power of identifying adaptive loci with *Bayescan* more than *PCAdapt* (Luu *et al.* 2017). This

latter, agreed with our data where we found admixture individuals in the whole SNPs

database. The presence of mixed individuals is supported by the high relative migration rates

estimated by Canales-Aguirre *et al.* (2022). Additionally, *BayeScan* could decrease power

with few populations (n <30 Aguirre-Liguori *et al.* 2020) and weak selection, when

populations deviate from the island model of migration and/or in cases of smaller sample size

(Foll & Gaggiotti 2008; Narum & Hess 2011).

Inside the GEA approach, *LFMM* detected the largest number of potentially adaptive

loci (n= 1,785 loci). Considering that *LFMM* identified 14.92% of the loci as adaptive, it is

likely that we will find false positives among them. The main reason is that geographic and

demographic processes can lead to patterns that mimic those observed due to selection. In

fact, de Villemereuil *et al*. (2014) found high rates of false discovery in some scenarios with

complex, hierarchical structures and polygenic selection; although showed that *LFMM*

provides the best compromise between detection power and error rates in situations with

complex hierarchical neutral genetic structure and polygenic selection. However, Frichot et

al. (2013) found that *LFMM* has low rates of false positives and negatives and that it performs

slightly better than *Bayenv* in detecting weak selection. Finally, Lotterhos & Whitlock (2015)

showed that *LFMM* is quite robust to various sampling designs and underlying demographic

models. The *MSOD* detected considerably fewer potentially adaptive loci (n=377 loci).

However, the main advantage of this program is that using Moran spectral randomization

(MSR) for significance testing, avoids the inflation of type I errors due to spatial

autocorrelation, whether resulting from gene flow or the spatial structure of the

environmental factors that induce selection (Wagner & Dray 2015; Wagner *et al.* 2017).

Moreover, spatial outlier detection worked best for scenarios of IBD, even in the presence of

large-scale spatial genetic structure (Wagner *et al.* 2017) as in this study. *RDA* detected the

fewest number of putative adaptive loci (n=83 loci) compared with *LFMM* and *MOSD*. The

main difference is the multivariate approach of *RDA*, while *LFMM* and *MSOD* are univariate,

meaning that they test one genetic marker at a time, and may also test only one environmental

predictor at a time (Rellstab *et al.* 2015); *RDA* identifying covarying loci associated with the

multivariate environmental variables (Legendre & Legendre 2012). The decrease in the

number of loci detected by *RDA* is because it restricts the number of environmental variables

tested based on the variance inflation factor. In the case of univariate programs, we tested six

environmental variables with four statistics for each but restricted to one statistic per

environmental variable for *RDA*, limiting the number of possible correlations. However,

simulated data showed that *RDA* has higher true-positive and lower false-positive rates

compared to *PCAdapt* and *LFMM* (Forester *et al.* 2018; Capblancq *et al.* 2018). Furthermore,

RDA has one important limitation: it models linear relationships among response and

predictors, meaning that nonlinear relationships will not be detected (Capblancq & Forester

2021). Finally, we did not detect any putative adaptive loci using the *BayeScEnv* and

Samβada programs. Considering that *BayeScEnv* produces fewer candidate loci than

BayeScan (de Villemereuil & Gaggiotti 2015), it is to be expected that if *BayeScan* did not

detect any loci *BayeScEnv* either. To date, we have not identified any studies that compare

the statistical power or control of FDR of *BayeScEnv* with respect to other programs.

Therefore, we cannot conclude whether *BayeScEnv* is more or less sensitive to certain

characteristics such as hierarchical population structure, isolation by distance, or strength of

selection. Due to the recentness of the *Samβada*, we did not identify comparative studies that

allow us to evaluate the advantages and disadvantages of this with respect to other GEA

approaches.

Continued testing and comparing PGD and GEA approaches is needed in simulation

frameworks that include more population structure, multiple selection surfaces and genetic

architectures that are more complex than the multilocus selection response modelled here.

Overall, we highlight the importance of combining PGD and GEA approaches that consider

multiple scenarios of adaptive variation, to diminish type I and II errors and avoid conclude

erroneously.

4.5 Implications for management and conservation

The 11,961 filtered loci genotyped for 202 individuals (the largest population genomic data set reported so far for *Eleginops maclovinus*) offers high resolution for creating management policies and decisions, with respect to patterns of adaptive genetic variation. Although neutral loci did not reveal a clear genetic structure among locations, analysis of 131 shared adaptive loci indicated genetic differentiation separating the populations into four groups: REL, CHE, CHA, and a fourth group comprising the remaining seven sampling locations. Note that all neutral pairwise F_{ST} values were close to zero, similar to showed by Canales-Aguirre *et al*. (2022) and also from fish and other marine species (Nayfa & Zenger 2016; Asaduzzaman *et al.* 2019b; a; O'Leary *et al.* 2021). The interpretation of extremely small F_{ST} is not straightforward (Waples & Gaggiotti 2006; Conover *et al.* 2006; Waples *et al.* 2008), meaning it is unclear what the implications of the patterns of weak neutral differentiation may be for demographic connectivity of subpopulations. Nevertheless, the signals of putatively adaptive genetic differentiation between the ten locations were significant. Our study demonstrates how the environment shapes the distribution of adaptive genetic variation across space for *E. maclovinus*; showing that REL, CHE and CHA retain adaptive characteristics different from the rest. This can have important implications for prioritizing sites for protection, where the level of adaptive genetic variation could be an indicator of the evolutionary resilience of populations (Bonin *et al.* 2007; Sgrò *et al.* 2011). Moreover, understanding the spatial distribution of putatively adaptive genetic variation can inform the selection of specific sites for protection within marine reserves to maintain the adaptive potential and evolutionary resilience of wild populations faced with environmental

change (von der Heyden 2017). The adaptive divergence reported here appears to be

underpinned by a variation in seawater temperature and oxygen concentrations in the

Northern Patagonia (Iriarte *et al.* 2014; Torres *et al.* 2014; Pérez-Santos *et al.* 2014; Silva &

Vargas 2014) and should affect traits relevant for the survival, growth and reproduction of *E.*

maclovinus. The complex patterns of population structure and putatively adaptive diversity

among *E. maclovinus* locations that we showed here are not recognized in current *E.*

maclovinus management units. To date, there is no management plan for *E. maclovinus* in

Chile, only regulations for the type of fishing gear used. Future research could use neutral

and adaptative datasets to determine management units using adaptative (e.g., population

adaptative index, adaptative score) and neutral (e.g., expected heterozygosity, local

differentiation) metrics as suggested by Xuereb et al. (2021). In the Chilean Northern

Patagonia, temperature and heat content are predicted to increase and freshwater river inputs

are predicted to decrease due to climate change (Aguayo *et al.* 2019). As such, the spatial

patterns of genetic variability observed in this study can be information conservation

planning decisions by identifying for protection populations containing higher levels of

segregating polymorphisms associated with environmental conditions (e.g., temperature and

salinity) that are prone to change in the future.

Literature cited

Aguayo R, León-Muñoz J, Vargas-Baecheler J *et al.* (2019) The glass half-empty: climate

change drives lower freshwater input in the coastal system of the Chilean Northern

Patagonia. *Climatic change*, **155**, 417–435.

Aguirre-Liguori JA, Luna-Sánchez JA, Gasca-Pineda J, Eguiarte LE (2020) Evaluation of

the Minimum Sampling Design for Population Genomic and Microsatellite Studies:

An Analysis Based on Wild Maize. *Frontiers in genetics*, **11**, 870.

Altshuler D, Pollara VJ, Cowles CR *et al.* (2000) An SNP map of the human genome

generated by reduced representation shotgun sequencing. *Nature*, **407**, 513–516.

Andrews KR, Good JM, Miller MR, Luikart G, Hohenlohe PA (2016) Harnessing the power

of RADseq for ecological and evolutionary genomics. *Nature reviews. Genetics*, **17**,

81–92.

Aracena C, Lange CB, Iriarte JL, Rebolledo L, Pantoja S (2011) Latitudinal patterns of export

production recorded in surface sediments of the Chilean Patagonian fjords (41--55 S)

as a response to water column productivity. *Continental shelf research*, **31**, 340–355.

Araneda C, Larraín MA, Hecht B, Narum S (2016) Adaptive genetic variation distinguishes

Chilean blue mussels *(Mitilus chilensis)* from different marine environments. *Ecology*

and evolution, **6**, 3632–3644.

Asaduzzaman M, Igarashi Y, Wahab MA *et al.* (2019a) Population Genomics of an

Anadromous Hilsa Shad *Tenualosa ilisha* Species across Its Diverse Migratory

Habitats: Discrimination by Fine-Scale Local Adaptation. *Genes*, **11**.

Asaduzzaman M, Wahab MA, Rahman MJ *et al.* (2019b) Fine-scale population structure and

ecotypes of anadromous Hilsa shad *(Tenualosa ilisha)* across complex aquatic

ecosystems revealed by NextRAD genotyping. *Scientific reports*, **9**, 16050.

Assis J, Tyberghein L, Bosch S (2018) Bio-ORACLE v2. 0: Extending marine data layers

for bioclimatic modelling. *Global Ecology and*.

Athauda S, Anderson T, de Nys R (2012) Effect of rearing water temperature on protandrous

sex inversion in cultured Asian Seabass *(Lates calcarifer)*. *General and comparative*

endocrinology, **175**, 416–423.

Babin PJ, Cerdà J, Lubzens E (2007) *The Fish Oocyte: From Basic Studies to*

Biotechnological Applications. Springer Science & Business Media.

Babin C, Gagnaire P-A, Pavey SA, Bernatchez L (2017) RAD-Seq Reveals Patterns of

Additive Polygenic Variation Caused by Spatially-Varying Selection in the American

Eel (Anguilla rostrata). *Genome biology and evolution*, **9**, 2974–2986.

Baird NA, Etter PD, Atwood TS *et al.* (2008) Rapid SNP discovery and genetic mapping

using sequenced RAD markers. *PloS one*, **3**, e3376.

Baroiller J-F, Guiguen Y, Fostier A (1999) Endocrine and environmental aspects of sex

differentiation in fish. *Cellular and molecular life sciences: CMLS*, **55**, 910–931.

Barrett RDH, Hoekstra HE (2011) Molecular spandrels: tests of adaptation at the genetic

level. *Nature reviews. Genetics*, **12**, 767–780.

Beaumont MA (2005) Adaptation and speciation: what can Fst tell us? *Trends in ecology &*

evolution, **20**, 435–440.

Beirão J, Lewis JA, Wringe BF, Purchase CF (2018) A novel sperm adaptation to
evolutionary constraints on reproduction: Pre-ejaculatory sperm activation in the
beach spawning capelin (Osmeridae). *Ecology and evolution*, **8**, 2343–2349.

Bekkevold D, Höjesjö J, Nielsen EE *et al.* (2020) Northern European *Salmo trutta (L.)*
populations are genetically divergent across geographical regions and environmental
gradients. *Evolutionary applications*, **13**, 400–416.

Benestan L, Quinn BK, Maaroufi H *et al.* (2016) Seascape genomics provides evidence for
thermal adaptation and current-mediated population structure in American lobster
(Homarus americanus). *Molecular ecology*, **25**, 5073–5092.

Benjamini Y, Hochberg Y (1995) Controlling the false discovery rate: A practical and
powerful approach to multiple testing. *Journal of the Royal Statistical Society*, **57**,
289–300.

Benoît HP, Pepin P, Brown JA (2000) Patterns of metamorphic age and length in marine
fishes, from individuals to taxa. *Canadian journal of fisheries and aquatic sciences.*
Journal canadien des sciences halieutiques et aquatiques, **57**, 856–869.

Bernatchez L (2016) On the maintenance of genetic variation and adaptation to
environmental change: considerations from population genomics in fishes. *Journal*
of fish biology, **89**, 2519–2556.

Blaxter (1992) The effect of temperature on larval fishes. *Netherlands Journal of Zoology*.

Bonin A, Nicole F, Pompanon F, Miaud C, Taberlet P (2007) Population adaptive index: a
new method to help measure intraspecific genetic diversity and prioritize populations

for conservation. *Conservation biology: the journal of the Society for Conservation*

Biology, **21**, 697–708.

Boulanger E, Benestan L, Guerin P-E *et al.* (2022) Climate differently influences the

genomic patterns of two sympatric marine fish species. *The Journal of animal*

ecology, **91**, 1180–1195.

Breckler J, Au K, Cheng J, Hasson T, Burnside B (2000) Novel myosin VI isoform is

abundantly expressed in retina. *Experimental eye research*, **70**, 121–134.

Brian JV, Harris CA, Runnalls TJ *et al.* (2008) Evidence of temperature-dependent effects

on the estrogenic response of fish: implications with regard to climate change. *The*

Science of the total environment, **397**, 72–81.

Brickle P, Laptikhovsky V, Arkhipkin A (2005) Reproductive strategy of a primitive

temperate notothenioid *Eleginops maclovinus*. *Journal of fish biology*, **66**, 1044–

1059.

Canales-Aguirre CB, Ferrada-Fuentes S, Galleguillos R *et al.* (2018) High genetic diversity

and low-population differentiation in the Patagonian sprat *(Sprattus fuegensis)* based

on mitochondrial DNA. *Mitochondrial DNA. Part A, DNA mapping, sequencing, and*

analysis, **29**, 1148–1155.

Canales-Aguirre CB, Ferrada-Fuentes S, Galleguillos R, Hernández CE (2016) Genetic

Structure in a Small Pelagic Fish Coincides with a Marine Protected Area: Seascape

Genetics in Patagonian Fjords. *PloS one*, **11**, e0160670.

Canales-Aguirre CB, Larson WA, McKinney GJ *et al.* (2022) Neutral and adaptive loci

reveal fine-scale population structure in *Eleginops maclovinus* from north Patagonia.

Ecology and evolution, **12**, e9343.

Capblancq T, Forester BR (2021) Redundancy analysis: A Swiss Army Knife for landscape

genomics. *Methods in ecology and evolution / British Ecological Society*.

Capblancq T, Luu K, Blum MGB, Bazin E (2018) Evaluation of redundancy analysis to

identify signatures of local adaptation. *Molecular ecology resources*, **18**, 1223–1233.

Catchen JM, Hohenlohe PA, Bernatchez L *et al.* (2017) Unbroken: RADseq remains a

powerful tool for understanding the genetics of adaptation in natural populations.

Molecular ecology resources, **17**, 362–365.

Cattell RB (1966) The Scree Test For The Number Of Factors. *Multivariate behavioral*

research, **1**, 245–276.

Cayuela H, Rougemont Q, Laporte M *et al.* (2020) Shared ancestral polymorphisms and

chromosomal rearrangements as potential drivers of local adaptation in a marine fish.

Molecular ecology, **29**, 2379–2398.

Ceballos SG, Lessa EP, Licandeo R, Fernández DA (2016) Genetic relationships between

Atlantic and Pacific populations of the notothenioid fish *Eleginops maclovinus*: the

footprints of Quaternary glaciations in Patagonia. *Heredity*, **116**, 372–377.

Ceballos SG, Lessa EP, Victorio MF, Fernández DA (2012) Phylogeography of the sub-

Antarctic notothenioid fish *Eleginops maclovinus*: evidence of population expansion.

Marine biology, **159**, 499–505.

Chen Z, Cheng C-HC, Zhang J *et al.* (2008) Transcriptomic and genomic evolution under

constant cold in Antarctic notothenioid fish. *Proceedings of the National Academy of*

Sciences of the United States of America, **105**, 12944–12949.

Chen L, Lu Y, Li W *et al.* (2019) The genomic basis for colonizing the freezing Southern

Ocean revealed by Antarctic toothfish and Patagonian robalo genomes. *GigaScience*,

8.

Cheng C-HC, Chen L, Near TJ, Jin Y (2003) Functional antifreeze glycoprotein genes in

temperate-water New Zealand nototheniid fish infer an Antarctic evolutionary origin.

Molecular biology and evolution, **20**, 1897–1908.

Conesa A, Götz S (2008) Blast2GO: A comprehensive suite for functional analysis in plant

genomics. *International journal of plant genomics*, **2008**, 619832.

Conover DO, Clarke LM, Munch SB, Wagner GN (2006) Spatial and temporal scales of

adaptive divergence in marine fishes and the implications for conservation. *Journal*

of fish biology, **69**, 21–47.

Coppola D, Abbruzzetti S, Nicoletti F *et al.* (2012) ATP regulation of the ligand-binding

properties in temperate and cold-adapted haemoglobins. X-ray structure and ligand-

binding kinetics in the sub-Antarctic fish *Eleginops maclovinus*. *Molecular*

bioSystems, **8**, 3295–3304.

Coppola D, Giordano D, Vergara A *et al.* (2010) The hemoglobins of sub-Antarctic fishes of

the suborder Notothenioidei. *Polar science*, **4**, 295–308.

Cui Y, Yin K, Zheng Y *et al.* (2021) Mixed plasticizers aggravated apoptosis by NOD2-

RIP2-NF-κB pathway in grass carp hepatocytes. *Journal of hazardous materials*, **402**,

123527.

Cure K, Thomas L, Hobbs J-PA, Fairclough DV, Kennington WJ (2017) Genomic signatures

of local adaptation reveal source-sink dynamics in a high gene flow fish species.

Scientific reports, **7**, 8618.

Cuvier G, Valenciennes A (1830) *Histoire naturelle des poissons*. F.-G. Levrault.

Czypionka T, Vargas CA, Silva N *et al.* (2011) Importance of mixotrophic nanoplankton in

Aysén Fjord (Southern Chile) during austral winter. *Continental shelf research*, **31**,

216–224.

Dabney A, Storey JD, Warmes GR (2010) *qvalue: Q-value estimation for false discovery*

rate control.

Dalongeville A, Benestan L, Mouillot D, Lobreaux S, Manel S (2018) Combining six genome

scan methods to detect candidate genes to salinity in the Mediterranean striped red

mullet *(Mullus surmuletus)*. *BMC genomics*, **19**, 1–13.

Davey JW, Cezard T, Fuentes-Utrilla P *et al.* (2013) Special features of RAD Sequencing

data: implications for genotyping. *Molecular ecology*, **22**, 3151–3164.

Dávila PM, Figueroa D, Müller E (2002) Freshwater input into the coastal ocean and its

relation with the salinity distribution off austral Chile (35–55°S). *Continental shelf*

research, **22**, 521–534.

Delgado ML, Górski K, Habit E, Ruzzante DE (2019) The effects of diadromy and its loss

on genomic divergence: The case of amphidromous *Galaxias maculatus* populations.

Molecular ecology, **28**, 5217–5231.

Devlin B, Roeder K (1999) Genomic control for association studies. *Biometrics*, **55**, 997–

1004.

Diopere E, Vandamme SG, Hablützel PI *et al.* (2017) Seascape genetics of a flatfish reveals

local selection under high levels of gene flow. *ICES journal of marine science:*

journal du conseil, **75**, 675–689.

Dong XY, Qin JG, Zhang XM (2011) Fish adaptation to oxygen variations in aquaculture

from hypoxia to hyperoxia. *African journal of fisheries and aquaculture*, **2**, 23.

Dorwart MR, Shcheynikov N, Yang D, Muallem S (2008) The solute carrier 26 family of

proteins in epithelial ion transport. *Physiology* , **23**, 104–114.

Dowling DK, Simmons LW (2009) Reactive oxygen species as universal constraints in life-

history evolution. *Proceedings of the Biological sciences / The Royal Society*, **276**,

1737–1745.

Dray S, Bauman D, Blanchet G *et al.* (2021) *adespatial: Multivariate Multiscale Spatial*

Analysis. Github.

Dray S, Legendre P, Peres-Neto PR (2006) Spatial modelling: a comprehensive framework

for principal coordinate analysis of neighbour matrices (PCNM). *Ecological*

modelling, **196**, 483–493.

Duruz S, Sevane N, Selmoni O *et al.* (2019) Rapid identification and interpretation of gene-
environment associations using the new R.SamBada landscape genomics pipeline.
Molecular ecology resources, **19**, 1355–1365.

Eastman JT (1993) *Antarctic Fish Biology: Evolution in a Unique Environment*. Academic
Press.

Evanno G, Regnaut S, Goudet J (2005) Detecting the number of clusters of individuals using
the software STRUCTURE: a simulation study. *Molecular ecology*, **14**, 2611–2620.

Eve AMJ, Smith JC (2017) Knockdown of Laminin gamma-3 (Lamc3) impairs motoneuron
guidance in the zebrafish embryo. *Wellcome open research*, **2**, 111.

Excoffier L, Lischer HEL (2010) Arlequin suite ver 3.5: a new series of programs to perform
population genetics analyses under Linux and Windows. *Molecular ecology*
resources, **10**, 564–567.

Fago A, Forest E, Weber RE (2001) Hemoglobin and subunit multiplicity in the rainbow
trout *(Oncorhynchus mykiss)* hemoglobin system. *Fish physiology and biochemistry*,
24, 335–342.

Feuerlein R, Weber R (1994) Rapid and simultaneus measurement of anodic and cathodic
haemoglobins and ATP and GTP concentrations in minute quantities of fish blood.
The Journal of experimental biology, **189**, 273–277.

Flanagan SP, Jones AG (2017) Constraints on the FST-Heterozygosity Outlier Approach.
The Journal of heredity, **108**, 561–573.

Flanagan SP, Rose E, Jones AG (2016) Population genomics reveals multiple drivers of

population differentiation in a sex-role-reversed pipefish. *Molecular ecology*, **25**,

5043–5072.

Foll M, Gaggiotti O (2008) A genome-scan method to identify selected loci appropriate for

both dominant and codominant markers: a Bayesian perspective. *Genetics*, **180**, 977–

993.

Forester BR, Lasky JR, Wagner HH, Urban DL (2018) Comparing methods for detecting

multilocus adaptation with multivariate genotype-environment associations.

Molecular ecology, **27**, 2215–2233.

Fox MD, Williams GJ, Johnson MD *et al.* (2018) Gradients in Primary Production Predict

Trophic Strategies of Mixotrophic Corals across Spatial Scales. *Current biology: CB*,

28, 3355-3363.e4.

Francis RM (2017) pophelper: an R package and web app to analyse and visualize population

structure. *Molecular ecology resources*, **17**, 27–32.

François O, Martins H, Caye K, Schoville SD (2016) Controlling false discoveries in genome

scans for selection. *Molecular ecology*, **25**, 454–469.

Frichot E, François O (2015) LEA: An R package for landscape and ecological association

studies. *Methods in ecology and evolution / British Ecological Society*, **6**, 925–929.

Frichot E, Mathieu F, Trouillon T, Bouchard G, François O (2014) Fast and efficient

estimation of individual ancestry coefficients. *Genetics*, **196**, 973–983.

Frichot E, Schoville SD, Bouchard G, François O (2013) Testing for associations between

loci and environmental gradients using latent factor mixed models. *Molecular biology*

and evolution, **30**, 1687–1699.

Gagliano M, McCormick MI, Meekan MG (2007) Temperature-induced shifts in selective

pressure at a critical developmental transition. *Oecologia*, **152**, 219–225.

Gagnaire P-A, Broquet T, Aurelle D *et al.* (2015) Using neutral , selected , and hitchhiker

loci to assess connectivity of marine populations in the genomic era. *Evolutionary*

applications, **8**, 769–786.

Gagnaire P-A, Gaggiotti OE (2016) Detecting polygenic selection in marine populations by

combining population genomics and quantitative genetics approaches. *Current*

zoology, **62**, 603–616.

Gillet C, Breton B, Mikolajczyk T (1996) Effects of GnRHa and pimozide treatments on the

timing of ovulation and on egg quality in Arctic charr (Salvelinus alpinus) at 5 and

10 °C. *Aquatic living resources*, **9**, 257–263.

Ginsburg AS (1963) Sperm-egg association and its relationship to the activation of the egg

in salmonid fishes. *Journal of embryology and experimental morphology*, **11**, 13–33.

González HE, Castro L, Daneri G *et al.* (2011) Seasonal plankton variability in Chilean

Patagonia fjords: Carbon flow through the pelagic food web of Aysen Fjord and

plankton dynamics in the Moraleda Channel basin. *Continental shelf research*, **31**,

225–243.

Gosztonyi AE, Ae G (1974) Edad y crecimiento del "Róbalo" *Eleginops maclovinus*

(Osteichthyes, Nototheniidae) en aguas de la ría Deseado y sus adyacencias.

Goudet J (2005) hierfstat, a package for r to compute and test hierarchical F-statistics.
Molecular ecology notes, **5**, 184–186.

Gove JM, McManus MA, Neuheimer AB *et al.* (2016) Near-island biological hotspots in
barren ocean basins. *Nature communications*, **7**, 10581.

Haasl RJ, Payseur BA (2016) Fifteen years of genomewide scans for selection: trends,
lessons and unaddressed genetic sources of complication. *Molecular ecology*, **25**, 5–
23.

Hair JF, Anderson RE, Tatham RL, and Black WC (1995) *Multivariate data analysis with*
readings. Prentice Hall, Englewood Cliffs, N.J.

Herberg S, Gert KR, Schleiffer A, Pauli A (2018) The Ly6/uPAR protein Bouncer is
necessary and sufficient for species-specific fertilization. *Science*, **361**, 1029–1033.

Herkenne S, Ek O, Zamberlan M *et al.* (2020) Developmental and Tumor Angiogenesis
Requires the Mitochondria-Shaping Protein Opa1. *Cell metabolism*, **31**, 987-1003.e8.

von der Heyden S (2017) Making evolutionary history count: biodiversity planning for coral
reef fishes and the conservation of evolutionary processes. *Coral reefs* , **36**, 183–194.

Hoban S, Kelley JL, Lotterhos KE *et al.* (2016) Finding the Genomic Basis of Local
Adaptation: Pitfalls, Practical Solutions, and Future Directions. *The American*
naturalist, **188**, 379–397.

Hoey JA, Pinsky ML (2018) Genomic signatures of environmental selection despite near-
panmixia in summer flounder. *Evolutionary applications*, **11**, 1732–1747.

Hokanson KEF (2011) Temperature Requirements of Some Percids and Adaptations to the
Seasonal Temperature Cycle. *Journal of the Fisheries Research Board of Canada*.

Holliday F (1969) 4 the effects of salinity on the eggs and larvae of teleosts. *Fish Physiology*,

1, 293–311.

Houde ED (1989) Comparative growth, mortality and energetic of marine fish larvae:

temperature an implied latitudinal effects. *Fishery bulletin* , 471–495.

Iriarte JL, González HE, Liu KK, Rivas C, Valenzuela C (2007) Spatial and temporal

variability of chlorophyll and primary productivity in surface waters of southern Chile

(41.5--43 S). *Estuarine, coastal and shelf science*, **74**, 471–480.

Iriarte JL, Pantoja S, Daneri G (2014) Oceanographic Processes in Chilean Fjords of

Patagonia: from small to large-scale studies. *Progress in Oceanography*, **129**, 1–7.

Islam Z, Kato A, Romero MF, Hirose S (2011) Identification and apical membrane

localization of an electrogenic Na^+/Ca^{2+} exchanger NCX2a likely to be involved in

renal Ca^{2+} excretion by seawater fish. *American journal of physiology. Regulatory,*

integrative and comparative physiology, **301**, R1427-39.

Jacob BG, Tapia FJ, Daneri G *et al.* (2014) Springtime size-fractionated primary production

across hydrographic and PAR-light gradients in Chilean Patagonia (41–50°S).

Progress in oceanography, **129**, 75–84.

Janhunen M, Piironen J, Peuhkuri N (2010) Parental effects on embryonic viability and

growth in Arctic charr *Salvelinus alpinus* at two incubation temperatures. *Journal of*

fish biology, **76**, 2558–2570.

Jensen FB, Fago A, Weber RE (1998) Hemoglobin structure and function. *Fish Physiology*,

17, 1–40.

Jensen FB, Weber RE (1985) Proton and oxygen equilibria, their anion sensitivities and

interrelationships in tench hemoglobin. *Molecular physiology*, **7**, 41–50.

Jiang J-L, Xu J, Ye L *et al.* (2020) Identification of differentially expressed genes in gills of

tiger puffer *(Takifugu rubripes)* in response to low-salinity stress. *Comparative*

biochemistry and physiology. Part B, Biochemistry & molecular biology, **243–244**,

110437.

Jombart T (2008) adegenet: a R package for the multivariate analysis of genetic markers.

Bioinformatics , **24**, 1403–1405.

Jombart T, Ahmed I (2011) adegenet 1.3-1: new tools for the analysis of genome-wide SNP

data. *Bioinformatics* , **27**, 3070–3071.

Jombart T, Devillard S, Balloux F (2010) Discriminant analysis of principal components: a

new method for the analysis of genetically structured populations. *BMC genetics*, **11**,

94.

Khananshvili D (2013) The SLC8 gene family of sodium-calcium exchangers (NCX) -

structure, function, and regulation in health and disease. *Molecular aspects of*

medicine, **34**, 220–235.

Kock K-H (1992) *Antarctic Fish and Fisheries*. Cambridge University Press.

Körner O, Kohno S, Schönenberger R *et al.* (2008) Water temperature and concomitant

waterborne ethinylestradiol exposure affects the vitellogenin expression in juvenile

brown trout *(Salmo trutta)*. *Aquatic toxicology* , **90**, 188–196.

Kültz D (2015) Physiological mechanisms used by fish to cope with salinity stress. *The*

Journal of experimental biology, **218**, 1907–1914.

Lattuca ME, Boy CC, Vanella FA, Barrantes ME, Fernández DA (2018) Thermal responses

of three native fishes from estuarine areas of the Beagle Channel, and their

implications for climate change. *Hydrobiologia*, **808**, 235–249.

Laverty G, Skadhauge E (2012) Adaptation of teleosts to very high salinity. *Comparative*

biochemistry and physiology. Part A, Molecular & integrative physiology, **163**, 1–6.

Legendre P, Legendre L (2012) *Numerical Ecology*. Elsevier.

Lehtonen TK, Kvarnemo C (2015) Density effects on fish egg survival and infections depend

on salinity. *Marine ecology progress series*, **540**, 183–191.

Lewontin RC, Krakauer J (1973) Distribution of gene frequency as a test of the theory of the

selective neutrality of polymorphisms. *Genetics*, **74**, 175–195.

Lex A, Gehlenborg N, Strobelt H, Vuillemot R, Pfister H (2014) UpSet: Visualization of

Intersecting Sets. *IEEE transactions on visualization and computer graphics*, **20**,

1983–1992.

Li Y-L, Xue D-X, Zhang B-D, Liu J-X (2019) Population Genomic Signatures of Genetic

Structure and Environmental Selection in the Catadromous Roughskin Sculpin

Trachidermus fasciatus. *Genome biology and evolution*, **11**, 1751–1764.

Li J, Zhang W, Zhou P *et al.* (2022) Selenium deficiency induced apoptosis via mitochondrial

pathway caused by Oxidative Stress in porcine gastric tissues. *Research in veterinary*

science, **144**, 142–148.

Liao B-K, Deng A-N, Chen S-C, Chou M-Y, Hwang P-P (2007) Expression and water

calcium dependence of calcium transporter isoforms in zebrafish gill mitochondrion-

rich cells. *BMC genomics*, **8**, 354.

Linford P, Pérez-Santos I, Montero P *et al.* (2023a) Oceanographic processes favoring

deoxygenation inside Patagonian fjords.

Linford P, Pérez-Santos I, Montero P *et al.* (2024) Oceanographic processes driving low-

oxygen conditions inside Patagonian fjords. *Biogeosciences*, **21**, 1433–1459.

Linford P, Pérez-Santos I, Montes I *et al.* (2023b) Recent deoxygenation of Patagonian fjord

subsurface waters connected to the Peru–Chile undercurrent and equatorial

subsurface water variability. *Global biogeochemical cycles*, **37**.

Lischer HEL, Excoffier L (2012) PGDSpider: an automated data conversion tool for

connecting population genetics and genomics programs. *Bioinformatics* , **28**, 298–

299.

Luu K, Bazin E, Blum MGB (2017) pcadapt: an R package to perform genome scans for

selection based on principal component analysis. *Molecular ecology resources*, **17**,

67–77.

Manel S, Perrier C, Pratlong M *et al.* (2016) Genomic resources and their influence on the

detection of the signal of positive selection in genome scans. *Molecular ecology*, **25**,

170–184.

Maselko J, Andrews KR, Hohenlohe PA (2020) Long-lived marine species may be resilient

to environmental variability through a temporal portfolio effect. *Ecology and*

evolution, **10**, 6435–6448.

McKinney GJ, Larson WA, Seeb LW, Seeb JE (2017a) RADseq provides unprecedented

insights into molecular ecology and evolutionary genetics: comment on Breaking

RAD by Lowry et al. (2016). *Molecular ecology resources*, **17**, 356–361.

McKinney G, McPhee MV, Pascal C, Seeb JE, Seeb LW (2020) Network Analysis of

Linkage Disequilibrium Reveals Genome Architecture in Chum Salmon. *G3* , **10**,

1553–1561.

McKinney GJ, Waples RK, Seeb LW, Seeb JE (2017b) Paralogs are revealed by proportion

of heterozygotes and deviations in read ratios in genotyping-by-sequencing data from

natural populations. *Molecular Ecology*, **17**, 656–669.

Miller MR, Dunham JP, Amores A, Cresko WA (2007) Rapid and cost-effective

polymorphism identification and genotyping using restriction site associated DNA

(RAD) markers. *Genome / National Research Council Canada = Genome / Conseil*

national de recherches Canada.

Montero P, Daneri G, González HE *et al.* (2011) Seasonal variability of primary production

in a fjord ecosystem of the Chilean Patagonia: Implications for the transfer of carbon

within pelagic food webs. *Continental shelf research*, **31**, 202–215.

Montero P, Daneri G, Tapia F, Iriarte JL, Crawford D (2017) Diatom blooms and primary

production in a channel ecosystem of central Patagonia. *Latin American journal of*

aquatic research, **45**, 999–1016.

Narum SR, Hess JE (2011) Comparison of F(ST) outlier tests for SNP loci under selection.

Molecular ecology resources, **11 Suppl 1**, 184–194.

Nayfa MG, Zenger KR (2016) Unravelling the effects of gene flow and selection in highly

connected populations of the silver-lip pearl oyster (Pinctada maxima). *Marine*

genomics, **28**, 99–106.

Nikinmaa M, Rees BB (2005) Oxygen-dependent gene expression in fishes. *American*

journal of physiology. Regulatory, integrative and comparative physiology, **288**,

R1079-90.

North AW (2001) Early life history strategies of notothenioids at South Georgia. *Journal of*

fish biology, **58**, 496–505.

O, Wai-sum, Chen H, Chow PH (2006) Male genital tract antioxidant enzymes—Their ability

to preserve sperm DNA integrity. *Molecular and cellular endocrinology*, **250**, 80–83.

Oksanen J, Guillaume F, Blanchet M *et al.* (2016) *vegan: Community Ecology Package*.

Okuzawa K, Kumakura N, Gen K *et al.* (2003) Effect of high water temperature on brain-

pituitary-gonad axis of the red seabream during its spawning season. *Fish physiology*

and biochemistry, **28**, 439–440.

O'Leary SJ, Hollenbeck CM, Vega RR, Portnoy DS (2021) Disentangling complex genomic

signals to understand population structure of an exploited, estuarine-dependent

flatfish. *Ecology and evolution*.

Oyarzún R, Muñoz JLP, Pontigo JP, Morera FJ, Vargas-Chacoff L (2018) Effects of

acclimation to high environmental temperatures on intermediary metabolism and

osmoregulation in the sub-Antarctic notothenioid *Eleginops maclovinus*. *Marine*

biology, **165**, 1–15.

Oyarzún R, Rojas JJ, Pontigo JP *et al.* (2021) Long-term effects of temperatures on the

physiological response of juveniles of the eurythermal sub-antarctic notothenioid

Eleginops maclovinus. *Aquaculture* , **530**, 1–11.

Pankhurst NW, King HR (2010) Temperature and salmonid reproduction: implications for

aquaculture. *Journal of fish biology*, **76**, 69–85.

Pantoja S, Iriarte J, Daneri G (2011) Oceanography of the Chilean Patagonia. *Continental*

shelf research, **31**, 149–153.

Paredes A, Montecino V (2011) Size diversity as an expression of phytoplankton community

structure and the identification of its patterns on the scale of fjords and channels.

Continental shelf research, **31**, 272–281.

Pauls SU, Nowak C, Bálint M, Pfenninger M (2013) The impact of global climate change on

genetic diversity within populations and species. *Molecular ecology*, **22**, 925–946.

Pauly D, Pullin RSV (1988) Hatching time in spherical, pelagic, marine fish eggs in response

to temperature and egg size. *Environmental biology of fishes*, **22**, 261–271.

Pembleton LW, Cogan NOI, Forster JW (2013) StAMPP: an R package for calculation of

genetic differentiation and structure of mixed-ploidy level populations. *Molecular*

ecology resources, **13**, 946–952.

Pequeño GH (1989) The geographical distribution and taxonomic arrangement of South

American notothenidae fishes (Osteichthyes, Notothenidae). *Bol. Soc. Biol.*

Concepción, Chile, **60**, 183–200.

Pérez-Santos I, Garcés-Vargas J, Schneider W *et al.* (2014) Double-diffusive layering and

mixing in Patagonian fjords. *Progress in oceanography*, **129**, 35–49.

Pickard GL (1971) Some physical oceanographic features of inlets of Chile. *Journal of the*

Fisheries Research Board of Canada, **28**, 1077–1106.

Pizarro G, Iriarte JL, Montecino V, Blanco JL, Guzmán L (2000) Distribución de la biomasa

fitoplanctónica y productividad primaria máxima de fiordos y canales australes (47–

50 S) en octubre 1996. *Ciencia y Tecnologia del Mar* .

Powers DA (1974) Structure, function, and molecular ecology of fish hemoglobins. *Annals*

of the New York Academy of Sciences, **241**, 472–490.

di Prisco G, Maresca B, Tota B (2012) *Biology of Antarctic Fish*. Springer Science &

Business Media.

di Prisco G, Tamburrini M (1992) The hemoglobins of marine and freshwater fish: the search

for correlations with physiological adaptation. *Comparative biochemistry and*

physiology. B, Comparative biochemistry, **102**, 661–671.

Pritchard JK, Stephens M, Donnelly P (2000) Inference of Population Structure Using

Multilocus Genotype Data. *Genetics*, **155**, 945–959.

Rabassa J, Coronato A, Martínez O (2011) Late Cenozoic glaciations in Patagonia and Tierra

del Fuego: an updated review. *Biological journal of the Linnean Society. Linnean*

Society of London, **103**, 316–335.

Rellstab C, Gugerli F, Eckert AJ, Hancock AM, Holderegger R (2015) A practical guide to

environmental association analysis in landscape genomics. *Molecular ecology*, **24**,

4348–4370.

Richards JG (2009) Chapter 10 Metabolic and Molecular Responses of Fish to Hypoxia. In:

Fish Physiology (eds Richards JG, Farrell AP, Brauner CJ), pp. 443–485. Academic

Press.

Rodrigo C (2008) Submarine topography in the Chilean North Patagonian channels. *Progress*

in the oceanographic knowledge of Chilean inner waters, from Puerto Montt to Cape

Horn. Comité Oceanográfico Nacional-Pontificia Universidad Católica de

Valparaíso, Valparaíso, 19–23.

Rombough PJ (1997) The effects of temperature on embryonic and larval development. In:

Global Warming: Implications for Freshwater and Marine Fish, pp. 177–224.

Cambridge University Press.

Rungruangsak-Torrissen K (2007) Digestive efficiency, growth and qualities of muscle and

oocyte in Atlantic salmon *(Salmo Salar L.)* fed on diets with krill meal as an

alternative protein source. *Journal of food biochemistry*, **31**, 509–540.

Rungruangsak-Torrissen K, Moss R, Andresen LH, Berg A, Waagbø R (2006) Different

expressions of trypsin and chymotrypsin in relation to growth in Atlantic salmon

(Salmo salar L.). *Fish physiology and biochemistry*, **32**, 7–23.

Schlitzer R (2002) Interactive analysis and visualization of geoscience data with Ocean Data

View. *Computers & geosciences*, **28**, 1211–1218.

Schneider W, Donoso D, Garcés-Vargas J, Escribano R (2017) Water-column cooling and

sea surface salinity increase in the upwelling region off central-south Chile driven by

a poleward displacement of the South Pacific High. *Progress in oceanography*, **151**,

38–48.

Schneider W, Pérez-Santos I, Ross L *et al.* (2014) On the hydrography of Puyuhuapi

Channel, Chilean Patagonia. *Progress in oceanography*, **129**, 8–18.

Scott SG, Pankhurst NW (1992) Interannual variation in the reproductive cycle of the New

Zealand snapper *Pagrus auratus* (Bloch & Schneider) (Sparidae). *Journal of fish*

biology, **41**, 685–696.

Segovia NI, González-Wevar CA, Haye PA (2020) Signatures of local adaptation in the

spatial genetic structure of the ascidian *Pyura chilensis* along the southeast Pacific

coast. *Scientific reports*, **10**, 14098.

Sgrò CM, Lowe AJ, Hoffmann AA (2011) Building evolutionary resilience for conserving

biodiversity under climate change. *Evolutionary applications*, **4**, 326–337.

Shimizu A (2003) Effect of photoperiod and temperature on gonadal activity and plasma

steroid levels in a reared strain of the mummichog *(Fundulus heteroclitus)* during

different phases of its annual reproductive cycle. *General and comparative*

endocrinology, **131**, 310–324.

Silva C, Andrade I, Yáñez E *et al.* (2016) Predicting habitat suitability and geographic

distribution of anchovy (Engraulis ringens) due to climate change in the coastal areas

off Chile. *Progress in oceanography*, **146**, 159–174.

Silva N, Vargas CA (2014) Hypoxia in Chilean Patagonian Fjords. *Progress in*

oceanography, **129**, 62–74.

Smith GD, Hopkins GR, Mohammadi S *et al.* (2015) Effects of temperature on embryonic

and early larval growth and development in the rough-skinned newt *(Taricha*

granulosa). *Journal of thermal biology*, **51**, 89–95.

Song X-B, Liu G, Liu F *et al.* (2017) Autophagy blockade and lysosomal membrane

permeabilization contribute to lead-induced nephrotoxicity in primary rat proximal

tubular cells. *Cell death & disease*, **8**, e2863.

Soria FN, Strüssmann CA, Miranda LA (2008) High water temperatures impair the

reproductive ability of the pejerrey fish *Odontesthes bonariensis*: effects on the

hypophyseal-gonadal axis. *Physiological and biochemical zoology: PBZ*, **81**, 898–

905.

Spanier KI, Leese F, Mayer C *et al.* (2010) Predator-induced defences in *Daphnia pulex*:

selection and evaluation of internal reference genes for gene expression studies with

real-time PCR. *BMC molecular biology*, **11**, 50.

Stucki S, Orozco-terWengel P, Forester BR *et al.* (2017) High performance computation of

landscape genomic models including local indicators of spatial association.

Molecular ecology resources, **17**, 1072–1089.

Sumpter JP, Jobling S (1995) Vitellogenesis as a biomarker for estrogenic contamination of

the aquatic environment. *Environmental health perspectives*, **103 Suppl 7**, 173–178.

Taylor CT, McElwain JC (2010) Ancient atmospheres and the evolution of oxygen sensing

via the hypoxia-inducible factor in metazoans. *Physiology* , **25**, 272–279.

Taylor KM, Nicholson RI (2003) The LZT proteins; the LIV-1 subfamily of zinc transporters.

Biochimica et biophysica acta, **1611**, 16–30.

Tian T, Zhao L, Zhang M, Zhao X, Meng A (2009) Both foxj1a and foxj1b are implicated in

left-right asymmetric development in zebrafish embryos. *Biochemical and*

biophysical research communications, **380**, 537–542.

Torres R, Silva N, Reid B, Frangopulos M (2014) Silicic acid enrichment of subantarctic

surface water from continental inputs along the Patagonian archipelago interior sea

(41--56 S). *Progress in oceanography*, **129**, 50–61.

Tveiten H, Solevag SE, Johnsen HK (2001) Holding temperature during the breeding season

influences final maturation and egg quality in common wolffish. *Journal of fish*

biology, **58**, 374–385.

Tyberghein L, Verbruggen H, Pauly K *et al.* (2012) Bio-ORACLE: a global environmental

dataset for marine species distribution modelling. *Global ecology and biogeography:*

a journal of macroecology, **21**, 272–281.

Van Der Kraak G, Pankhurst NW (1997) Temperature effects on the reproductive

performance of fish. In: *Global Warming: Implications for Freshwater and Marine*

Fish, pp. 159–176. Cambridge University Press.

Vanella FA, Boy CC, Fernández DA (2012) Temperature effects on growing, feeding, and

swimming energetics in the Patagonian blennie *Eleginops maclovinus* (Pisces:

Perciformes). *Polar biology*, **35**, 1861–1868.

Vargas-Chacoff L, Moneva F, Oyarzún R *et al.* (2014) Environmental salinity-modified

osmoregulatory response in the sub-Antarctic notothenioid fish *Eleginops*

maclovinus. *Polar biology*, **37**, 1235–1245.

Vargas-Chacoff L, Moneva F, Oyarzún R *et al.* (2015a) Metabolic responses to salinity

changes in the subantarctic notothenioid teleost *Eleginops maclovinus*. *Polar biology*,

39, 1297–1308.

Vargas-Chacoff L, Saavedra E, Oyarzún R *et al.* (2015b) Effects on the metabolism, growth,

digestive capacity and osmoregulation of juvenile of Sub-Antarctic Notothenioid fish

Eleginops maclovinus acclimated at different salinities. *Fish physiology and*

biochemistry, **41**, 1369–1381.

Venables WN, Ripley BD (2013) *Modern Applied Statistics with S-PLUS*. Springer Science

& Business Media.

de Villemereuil P, Frichot É, Bazin É, François O, Gaggiotti OE (2014) Genome scan

methods against more complex models: when and how much should we trust them?

Molecular ecology, **23**, 2006–2019.

de Villemereuil P, Gaggiotti OE (2015) A new F ST -based method to uncover local

adaptation using environmental variables (RB O'Hara, Ed,). *Methods in ecology and*

evolution / British Ecological Society, **6**, 1248–1258.

Wagner HH, Chávez-Pesqueira M, Forester BR (2017) Spatial detection of outlier loci with

Moran eigenvector maps. *Molecular ecology resources*, **17**, 1122–1135.

Wagner HH, Dray S (2015) Generating spatially constrained null models for irregularly

spaced data using Moran spectral randomization methods. *Methods in ecology and*

evolution / British Ecological Society, **6**, 1169–1178.

Wald A (1943) Tests of Statistical Hypotheses Concerning Several Parameters When the

Number of Observations is Large. *Transactions of the American Mathematical*

Society, **54**, 426–482.

Wang L, Wan ZY, Lim HS, Yue GH (2015) Genetic heterogeneity and local adaptation of

Asian seabass across Indonesian Archipelago revealed with gene-associated SNP

markers. *Fisheries research*, **170**, 205–211.

Wang Z, Wu J, Hu Z *et al.* (2020) Dexmedetomidine Alleviates Lipopolysaccharide-Induced

Acute Kidney Injury by Inhibiting p75NTR-Mediated Oxidative Stress and

Apoptosis. *Oxidative medicine and cellular longevity*, **2020**, 5454210.

Waples RS, Gaggiotti O (2006) What is a population? An empirical evaluation of some

genetic methods for identifying the number of gene pools and their degree of

connectivity. *Molecular ecology*.

Waples RS, Punt AE, Cope JM (2008) Integrating genetic data into management of marine

resources: how can we do it better? *Fish and fisheries* .

Weber (1990) Functional significance and structural basis of multiple hemoglobins with

special reference to ectothermic vertebrates. *Animal nutrition and transport*

processes.

Weber RE, Fago A, Val AL *et al.* (2000) Isohemoglobin differentiation in the bimodal-

breathing amazon catfish *Hoplosternum littorale*. *The Journal of biological*

chemistry, **275**, 17297–17305.

Weir BS, Cockerham CC (1984) Estimating F-Statistics for the Analysis of Population

Structure. *Evolution; international journal of organic evolution*, **38**, 1358–1370.

Whitlock MC, Lotterhos KE (2015) Reliable Detection of Loci Responsible for Local

Adaptation: Inference of a Null Model through Trimming the Distribution of F(ST).

The American naturalist, **186 Suppl 1**, S24-36.

Wigginton JE, Cutler DJ, Abecasis GR (2005) A note on exact tests of Hardy-Weinberg

equilibrium. *American journal of human genetics*, **76**, 887–893.

Willette DA, Allendorf FW, Barber PH *et al.* (2014) So, you want to use next-generation

sequencing in marine systems? Insight from the Pan-Pacific Advanced Studies

Institute. *Bulletin of marine science*, **90**, 79–122.

Wu RSS (2009) Chapter 3 Effects of Hypoxia on Fish Reproduction and Development. In:

Fish Physiology (eds Richards JG, Farrell AP, Brauner CJ), pp. 79–141. Academic

Press.

Xu Q, Cheng C-HC, Hu P *et al.* (2008) Adaptive evolution of hepcidin genes in antarctic

notothenioid fishes. *Molecular biology and evolution*, **25**, 1099–1112.

Xuereb A, D'Aloia CC, Andrello M, Bernatchez L, Fortin M-J (2021) Incorporating

putatively neutral and adaptive genomic data into marine conservation planning.

Conservation biology: the journal of the Society for Conservation Biology, **35**, 909–

920.

Yanagimachi, Harumi, Matsubara (2017) Chemical and physical guidance of fish

spermatozoa into the egg through the micropyle. *Biology of blood and marrow

transplantation: journal of the American Society for Blood and Marrow

Transplantation*.

Yilmaz O, Patinote A, Nguyen T *et al.* (2019) Genome editing reveals reproductive and

developmental dependencies on specific types of vitellogenin in zebrafish *(Danio

rerio)*. *Molecular reproduction and development*, **86**, 1168–1188.

Yilmaz O, Patinote A, Nguyen T, Bobe J (2018) Multiple vitellogenins in zebrafish *(Danio*

rerio): quantitative inventory of genes, transcripts and proteins, and relation to egg

quality. *Fish physiology and biochemistry*, **44**, 1509–1525.

Yingjie K, Haihong Y, Lingwei C *et al.* (2019) Apoptosis repressor with caspase recruitment

domain deficiency accelerates ischemia/reperfusion (I/R)-induced acute kidney

injury by suppressing inflammation and apoptosis: The role of AKT/mTOR

signaling. *Biomedicine & pharmacotherapy = Biomedecine & pharmacotherapie*,

112, 108681.

Zhao L, Tan J, Li D *et al.* (2019) SLC39A6/ZIP6 is essential for zinc homeostasis and T-cell

development in zebrafish. *Biochemical and biophysical research communications*,

511, 896–902.

Zuur AF, Ieno EN, Elphick CS (2010) A protocol for data exploration to avoid common

statistical problems. *Methods in ecology and evolution / British Ecological Society*,

1, 3–1