




ORIGINAL ARTICLE OPEN ACCESS

Defining Conservation Units in a Highly Diverse Species: A Case on Arctic Charr

Sam Fenton¹  | Colin W. Bean^{1,2}  | Samuel A. M. Martin³  | Samuel J. Poultney⁴ | Antony Smith⁵  | Elvira de Eyto⁶  | Kathryn R. Elmer¹  | Colin E. Adams^{1,7} 

¹School of Biodiversity, One Health & Veterinary Medicine, College of Medical, Veterinary & Life Sciences, University of Glasgow, Glasgow, UK | ²NatureScot, Clydebank, UK | ³School of Biological Sciences, University of Aberdeen, Aberdeen, UK | ⁴Dee District Salmon Fishery Board, Aboyne, UK | ⁵Molecular Ecology and Evolution at Bangor (MEEB), School of Natural Sciences, Bangor University, Bangor, Gwynedd, UK | ⁶Marine Institute, Newport, Ireland | ⁷Scottish Centre for Ecology and the Natural Environment, University of Glasgow, Rowardennan, UK

Correspondence: Kathryn R. Elmer (kathryn.elmer@glasgow.ac.uk)

Received: 9 May 2025 | **Revised:** 9 December 2025 | **Accepted:** 11 December 2025

Samuel J. Poultney—<https://www.researchgate.net/profile/Samuel-Poultney>.

Keywords: conservation units | evolutionarily significant units | fish | genomic vulnerability | management units | population genomics

ABSTRACT

Defining appropriate conservation units is crucial to the protection and management of biodiversity. These delineations deliver further benefit when they include assessments of population vulnerability to extinction from pressures such as climate change. However, delineations and vulnerability assessments are particularly difficult within highly diverse species, such as the salmonid fish Arctic charr (*Salvelinus alpinus*), that show extensive phenotypic and genetic variation within and across locations, variable and complex life histories and broad geographic distributions. As yet, the nature and scope of Arctic charr diversity has not been characterised at the scale needed to delineate key conservation units in Scotland. To identify evolutionarily significant and vulnerable populations to prioritise for conservation, we conducted a genomic study of Arctic charr populations across Britain and Ireland with a focus on Scottish populations ($N = 64$ populations; 24,878 SNPs; 410 individuals). We found that most lake populations represented distinct genetic clusters, with limited gene flow between them and resulting in substantial genetic differentiation. Higher level groupings of genetic similarity across catchments likely reflect historic anadromy and migration, with populations primarily grouping east or west of the central watershed divide in Scotland. Analysing genetic offset, also known as genomic vulnerability, we identified strong inverse correlations between genetic vulnerability and latitude and distance to the sea, suggesting that more southern and more inland populations are more vulnerable to the effects of climate change. Additionally, patterns of vulnerability across several additional metrics identified other populations that may be at higher risk of loss. We further used our genetic data, along with phenotypic and geographic information, to identify populations of greatest evolutionary significance. This highlighted that the most important ones to protect are those in locations with multiple ecotypes, a key facet of functional Arctic charr biodiversity, and populations that are the only ones in their Hydrometric Area.

1 | Introduction

The management and protection of biodiversity requires both the delineation of appropriate units of conservation and an assessment of their vulnerability to pressures such as climate

change. While conservation action is frequently directed at the species level, an increasing body of work has demonstrated the importance of intraspecific, within-species, diversity to ecological versatility, ecosystem recovery and adaptation to changing environments (Le Coeur et al. 2021; Reusch

This is an open access article under the terms of the [Creative Commons Attribution](https://creativecommons.org/licenses/by/4.0/) License, which permits use, distribution and reproduction in any medium, provided the original work is properly cited.

© 2025 The Author(s). *Evolutionary Applications* published by John Wiley & Sons Ltd.

et al. 2005; Des Roches et al. 2018; Schindler et al. 2010). For many species, particularly highly diverse ones, conservation management at a species level is insufficient for protecting vital variation and it is becoming increasingly common practice to define intraspecific conservation units (Forester et al. 2022; Mee et al. 2015). However, there are many different functional definitions of what constitutes a biologically meaningful conservation unit, and the criteria used to define these units can drastically impact the assessment of conservation importance (Mee et al. 2015; Minter et al. 2021; Musmann et al. 2020). For example, the emphasis on morphological characteristics in the Kottelat and Freyhof (2007) review of the freshwater fishes of Europe led to a number of highly debated species classifications (Adams and Maitland 2007; Barthelemy et al. 2023; Crotti et al. 2020; Denys 2021; Etheridge et al. 2012; Gratton et al. 2014; Valavidis et al. 2019). As such, an approach is needed to delineate appropriate units of conservation based on genetics rather than inconsistent morphological traits, particularly on the intraspecific level.

With the aim of conserving the extant diversity of a species and protecting its adaptive potential (i.e., future biodiversity), definitions of intraspecific conservation units focus on identifying demographically independent groups, or groups showing high genetic differentiation as a proxy for demographic independence (Allendorf et al. 2022; Mable 2019). Two of the most commonly applied conservation delineations are Evolutionarily Significant Units (ESUs) and Management Units (MUs) (Allendorf et al. 2022; Funk et al. 2012; Moritz 1994). Broadly, an ESU can be defined as a demographically independent population, or group of populations, that show notable adaptive differentiation or functional diversity for the species. However, a number of variations of this definition exist, with some placing emphasis on contemporary patterns of diversity while others focus more on distinct historical and evolutionary lineages (Casacci et al. 2014; Coates et al. 2018; De Guia and Saitoh 2007; Moritz 1994; Waples 1991). In contrast, MUs are generally defined as demographically independent populations or groups of populations without the element of adaptive differentiation used for ESUs (Funk et al. 2012; Moritz 1994). Many studies define these units hierarchically, with a species consisting of several larger ESUs that themselves contain multiple MUs (Funk et al. 2012; Shaney et al. 2020). However, this is not always the case; a species may comprise a single ESU, an ESU may contain a single MU within it, for example, when an ESU consists of a sole population, and MUs may be delineated first with the candidacy of ESU status at different hierarchies then evaluated (Casacci et al. 2014; Forester et al. 2022).

Several data types, for example, genetic, phenotypic and geographic data, should be combined where possible when defining units of conservation (Hoelzel 2023). Information from different data types can provide important evidence that may reinforce the conclusion from another; this idea of concordance was a key part of the original definition of an ESU (Ryder 1986). In addition, different types of data can provide alternative insights into what is evolutionarily significant, for example, when high morphological differentiation is not matched by high genetic differentiation (de Apolônio Silva Oliveira et al. 2017; Funk et al. 2012). Nonetheless, many studies defining intraspecific conservation units have used a

criterion based solely on genetic data (Robertson et al. 2014). The advent of large-scale genomic datasets that use both neutral and adaptive loci has increased capacity to delineate conservation units more precisely and with greater biological depth (Allendorf et al. 2022; Funk et al. 2012). How conservation units are delineated is highly susceptible to whether patterns of differentiation are considered at a genome-wide scale or restricted to single or several adaptive genes (Waples and Lindley 2018; Waples et al. 2022).

Beyond the delineation of conservation units, another important requirement of conservation management is the need to identify habitats and sites that may require urgent conservation action or monitoring because they are more vulnerable to loss (Campbell et al. 2017; Sgrò et al. 2011). This can involve investigating both the population for conservation and its supporting habitat. For example, small habitats with lower carrying capacity will limit the size of a population and potentially its overall diversity, restricting its ability to respond to change (Fischer and Lindenmayer 2007).

Increasing temperatures are a threat to many species, with increased water temperature shown to affect the development and morphology of freshwater species (Campbell et al. 2021; Esin et al. 2021; Muir et al. 2022). The threat levels are magnified for populations in smaller lakes or streams which may have less refuge from high water temperatures when compared with populations in larger habitats (Campbell et al. 2017; Isaak and Young 2023). Thus, using the combination of population genetics and environmental information to identify the level of the threat to a population from various factors can help identify not just vulnerable species but also vulnerable sites for that species to prioritise for conservation action.

A species that aptly demonstrates the complexity of defining units of conservation is the Arctic charr (*Salvelinus alpinus*), a salmonid fish. Across its Holarctic range, the species exhibits very high variation in the expression of phenotypic traits and high levels of genetic variation (Gordeeva et al. 2015; Jonsson and Jonsson 2001; Klemetsen 2010; Maitland and Adams 2018; Recknagel et al. 2017; Wilson et al. 2004). There are a number of lakes that contain multiple distinct populations, known as ecotypes, diverged to specialise in different trophic niches and living in sympatry (Elmer 2016; Klemetsen 2010; Maitland and Adams 2018). These ecotypes are often distinguished by differences in their diet and foraging tactics but can be differentiated in other characters such as morphology and spawning time (Adams et al. 1998; Doenz et al. 2019; Garduño-Paz et al. 2012; Hooker et al. 2016; Jacobs et al. 2020; Kess et al. 2021; Malmquist et al. 1992; Skúlason et al. 1996). The presence of distinct sympatric ecotypes tends to be associated with local environmental features such as lake size and climate (Blain et al. 2025; Fenton, Bean, et al. 2024; Tiddy et al. 2024). The extent of divergence between co-existing sympatric ecotypes varies by location, and each represents an important example of the functional diversity of the species (Adams et al. 2007; Adams and Huntingford 2004; Fenton, Jacobs, et al. 2024; Gordeeva et al. 2015; Moccetti et al. 2019; Schluter 1996). Many of these instances of ecotype divergence are the result of secondary contact and admixture of multiple lineages that colonised post-glaciation, while others represent

recent sympatric divergences from a single lineage and therefore show lower levels of genetic differentiation but frequently clear differentiation in phenotypes (Garduño-Paz et al. 2012; Jacobs et al. 2020; Kettle-White 2001).

The high intraspecific diversity of Arctic charr has resulted in both the species and their habitats being designated as having high conservation, natural heritage and scientific value for many countries, including the UK (Bean et al. 2018). In Britain and Ireland, the species is not anadromous, remaining in freshwater throughout the life cycle (Finstad and Hein 2012; Jørgensen and Johnsen 2014; Maitland and Adams 2018), thereby inhibiting contemporary gene flow between populations in unconnected river catchments. This effect is compounded by the existence of multiple lakes which harbour sympatric and independently evolved ecotypes with genetic, phenotypic and ecological distinctiveness (Fenton, Jacobs, et al. 2024; Hooker et al. 2016; Jacobs et al. 2020). Given the natural heritage value and the geographic isolation of populations, there is a particular need to identify evolutionary significant and/or vulnerable sites for Arctic charr for conservation purposes. Arctic charr is listed as one of the priority fish species under the UK Biodiversity Framework (BRIG 2007), in recognition of the need for conservation action to protect the species and the substantial diversity seen within it. Currently, the conservation policy body for Scotland focuses both on preservation of habitats, such as through Sites of Special Scientific Interest (SSSI), Special Area of Conservation (SAC) and National Nature Reserve (NNR) sites—and on genetic conservation units (GCU) (Cavers et al. 2022). Recent guidelines for selection of SSSIs in Great Britain make a strong case for selection and protection of sites supporting within-species diversity and seeks to preserve the habitats giving rise to this diversity (Bean et al. 2018). There is also a concerted movement towards the protection of genetic diversity and future adaptive potential of species, particularly those known to be threatened, both in the UK and internationally in response to Aichi Targets (2011–2020) and the Kunming-Montreal framework (2022–2030) (Department of Agriculture, Environment and Rural Affairs (DAERA), et al. 2025; Hoban et al. 2025; Hollingsworth et al. 2019). However, given that management resources are limited, it is important to define populations, or groups of populations, that are arguably higher priority for protection than others.

To delineate conservation units and identify vulnerable populations, we conducted a national scale genomic study of Arctic charr in Scotland, contextualised across Britain and Ireland. We evaluated multiple approaches to determine populations for conservation action using a genome-wide dataset of SNPs. We analysed overall patterns of genome-wide genetic structuring and phylogenetic analysis across all populations and catchments. Further, we examined patterns of variation in putatively adaptive SNPs associated with lake environment. We inferred populations that we predict may be of higher risk to loss in the future, via explicit vulnerability metrics such as genetic offset (genomic vulnerability). We then used our analyses in tandem with phenotypic and ecological knowledge and geographical information, to delineate ESUs and MUs within the species and highlight populations of significance to protect.

2 | Materials and Methods

2.1 | Sample Collection

Arctic charr were sampled from 49 Scottish lakes, three Irish lakes, two Welsh lakes and one English lake (Figure 1) between 1997 and 2021 using 30 m x 1.5 m Nordic multi-panel mesh survey gill nets (CEN 2015) except at Llyn Padarn where fry traps were used to catch fry samples. Sample numbers and locations for each lake are indicated in Table 1. Eight of the sampled lakes in Scotland contain multiple sympatric ecotypes (Adams et al. 1998; Fraser et al. 1998; Jacobs et al. 2020). The dataset includes three instances of parapatric divergences (two neighbouring lakes from the same river system supporting two ecotypes): Shin-Merkland, Doine-Lubnaig and Stack-More (Adams et al. 2006; Adams, Bean, et al. 2007). Each ecotype was considered its own population in all steps of data generation and analysis unless otherwise specified. Therefore, in this paper, we use the term population to refer to all Arctic charr individuals in the same lake of origin or a group of individuals that share morphological/ecological traits within a lake. Lochs Maree and Stack are both known to contain multiple ecotypes (Adams et al. 2008); however, these were not sampled explicitly and so each of those lakes was considered as one population for the purposes of this study. In total, we had samples from 64 different Arctic charr populations covering 26 of the 27 Scottish Hydrometric Areas known to contain extant Arctic charr populations (Maitland and Adams 2018). The dataset included 10 individuals per population for each unimodal population and five individuals per ecotype in multiple ecotype lakes, wherever possible (Table 1). Hydrometric Areas (HAs) are defined as single river catchments or several contiguous river catchments which are designated across the United Kingdom (National River Flow Archive 2014).

2.2 | DNA Extraction and ddRADseq

DNA was extracted from fin clips, muscle, gill, head kidney, or from whole fry tissue using the Mackerey-Nagel NucleoSpin DNA Extraction Kit following manufacture recommendations. DNA quality and quantity were assessed using agarose gel electrophoresis and the Qubit Fluorometer with Life Technologies dsDNA BR assay. Four libraries of 90 samples plus five technical replicates were prepared using a ddRADseq protocol for Illumina, modified from Recknagel et al. (2015) and following Jacobs et al. (2020). Briefly, double restriction enzyme digestion was performed with *MspI* and *PstI*. A size selection range of 145–295 bp was isolated using Pippin Prep. Final RAD-tag enrichment was performed over 10 PCR amplification cycles. Paired end 75 bp sequencing was performed on the Illumina NextSeq500 platform at Glasgow Polyomics to generate 450M reads per library (target of 4M reads per individual and including 10% PhiX).

2.3 | Processing of ddRADseq Data

Raw sequence quality was assessed using *FastQC v0.11.9* (<https://www.bioinformatics.babraham.ac.uk/projects/fastqc/>). The *process_radtags* pipeline in *Stacks v2.60* was used

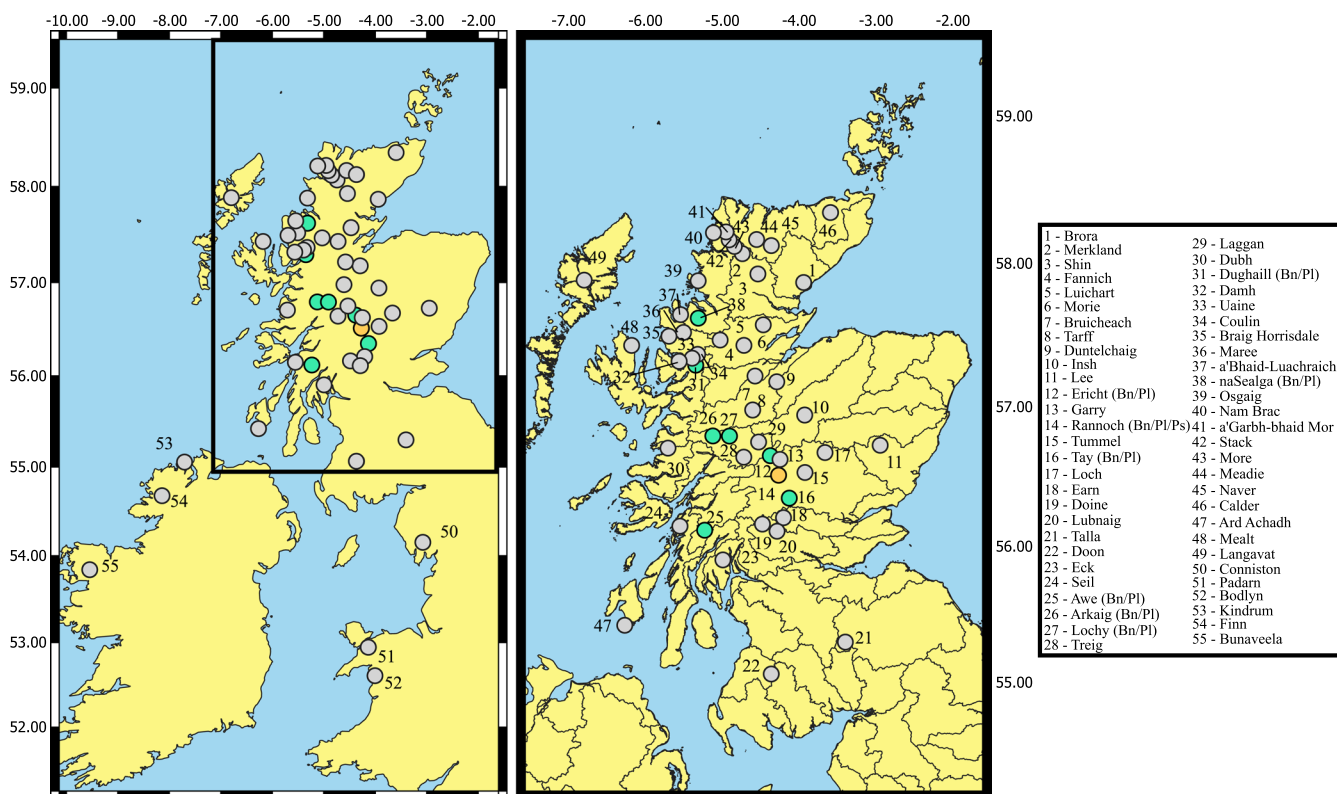


FIGURE 1 | Map showing the location of the populations from the British Isles included in this study. Lake origin is indicated on the map in ascending number order following Hydrometric Area order. Lakes with multiple ecotypes are indicated by Bn/Pl/Ps delineators in the legend and with green (bimodal) and oranges (trimodal) colours respectively. Bn refers to benthivore, Pl refers to planktivore and Ps refers to piscivore. Hydrometric Area boundaries are indicated on the Scotland specific panel on the right.

for demultiplexing raw sequence data (Rochette et al. 2019). To prevent a drop in sequence quality affecting mapping, the first 6 bp of the second read were removed using *Trimmomatic v0.39* (Bolger et al. 2014). New data were combined with previously generated data for eight Scottish lakes (Awe, Dughail, na Sealga, Tay, Lubnaig, Merkland, Uaine, Eck) from Jacobs et al. (2020) (ddRADseq NCBI short read bioproject: PRJNA607173). Mapping was performed to the annotated *Salvelinus* sp. genome (ASM291031v2) using *bwa mem v0.7.16* using a seed length of 20 (Li 2013). Reads with a mapping quality <20 were filtered out using *samtools v1.15* (Danecek et al. 2021). Ten individuals with <1M mapped reads were removed due to the high amounts of missing data. The *gstacks* module of *Stacks* was run with the `-gt-alpha` set to 0.01 to ensure called genotypes were reliable when RAD-loci building. SNPs were retained in the populations module of *Stacks* if they met the following criteria: present in 66% of all individuals in each population and across all populations, a minimum minor allele frequency of 0.01, maximum observed heterozygosity of 0.5 with only the first SNP per locus retained. We used the script *filter_hwe_by_pop.pl* (<https://github.com/jpuritz/dDocent>) to filter any SNPs that were out of Hardy–Weinberg Equilibrium within populations, although no such SNPs were detected. *VCFtools v0.1.16* was used to identify and remove any SNPs with read depth <10× or >30× using the 2× average depth rule to determine allowed maximum depth (Crotti et al. 2021; Danecek et al. 2011) with our average depth being ~15×.

2.4 | Genotyping Error

We used an R script published by Mastretta-Yanes et al. (2015) to estimate the rate of genotyping error using 20 replicated individuals (R Core Team 2021). Overall error rate was calculated to be 2.6%. The R script was then adapted to identify particular SNPs that showed a high rate of mismatching across technical replicates. SNPs that mismatched in more than 1 replicate pair (i.e., >5% genotyping error rate) were removed from the dataset for all individuals. Genotyping error was then recalculated from technical replicates and resulted in a mean error rate of 1.4%. Technical replicate samples were then removed from the dataset, retaining the replicate with the higher coverage. After filtering we had 24,878 SNPs and 410 individuals covering 64 different populations. While an upper limit of 33% missing data was allowed, average missing data per SNP was just over 10%, with roughly 4600 SNPs showing 20% or more missing data.

2.5 | Summary Statistics

Pairwise F_{ST} , observed and expected heterozygosity, number of private alleles and F_{IS} were generated for all populations and loci using the `-fstats` parameter in the *populations* module of *Stacks*. Population-specific F_{ST} was calculated using the *betas()* function in the *adegenet R package v2.1.10* (Jombart 2008; Kitada et al. 2021).

TABLE 1 | Overview of each of our 64 populations with summary statistics describing genomic diversity.

Pop ID	Latitude	Longitude	Sampling year	N	Pa	Ho	He	π	FST
a'Bhaid-Luachraich	57.81	-5.55	2012	10	32	0.093	0.075	0.080	0.512
a'Garbh-bhaid Mor	58.39	-4.95	2021	10	80	0.065	0.048	0.051	0.688
Ard Achadh	55.36	-6.16	2021	10	164	0.048	0.039	0.042	0.741
Arkaig (Benthivore)	56.97	-5.14	2007	4	0	0.116	0.094	0.110	0.329
Arkaig (Planktivore)	56.97	-5.14	2007	5	0	0.116	0.098	0.110	0.327
Awe (Benthivore)	56.30	-5.23	1997	5	0	0.119	0.108	0.125	0.245
Awe (Planktivore)	56.30	-5.23	1997	5	0	0.121	0.103	0.119	0.279
Bodlyn	52.80	-4.01	2002	2	0	0.112	0.089	0.128	0.210
Braig horrisdale	57.67	-5.67	2016	9	75	0.067	0.056	0.060	0.633
Brora	58.05	-3.95	2010	4	1	0.095	0.074	0.088	0.461
Bruicheach	57.39	-4.58	2009	8	24	0.075	0.059	0.065	0.601
Bunaveela	54.02	-9.54	2001	10	127	0.054	0.044	0.047	0.714
Calder	58.52	-3.59	2021	3	0	0.108	0.084	0.102	0.377
Conniston	54.34	-3.07	2005	1	0	0.129	0.065	0.129	0.165
Coulin	57.55	-5.35	1997	9	0	0.079	0.066	0.071	0.566
Damh	57.50	-5.57	2008	2	0	0.112	0.069	0.104	0.355
Doine	56.34	-4.48	1997	4	0	0.118	0.088	0.106	0.353
Doon	55.24	-4.37	1997	6	0	0.112	0.090	0.100	0.387
Dubh	56.88	-5.71	2021	5	12	0.124	0.091	0.108	0.339
Dughail (Benthivore)	57.47	-5.35	2013/2014	5	2	0.102	0.087	0.098	0.401
Dughail (Planktivore)	57.47	-5.35	2013/2014	5	0	0.105	0.093	0.105	0.361
Duntelchaig	57.35	-4.29	2010	4	0	0.122	0.089	0.113	0.305
Earn	56.38	-4.23	2016	9	27	0.102	0.083	0.089	0.453
Eck	56.08	-4.99	1997	10	59	0.060	0.049	0.057	0.653
Ericht (Benthivore)	56.82	-4.39	2007	4	1	0.127	0.106	0.126	0.229
Ericht (Planktivore)	56.82	-4.39	2007	5	0	0.145	0.121	0.136	0.166
Fannich	57.64	-4.96	2007	9	2	0.068	0.056	0.060	0.634
Finn	54.86	-8.14	2012	1	0	0.051	0.025	0.051	0.679
Garry	56.80	-4.25	2007	8	2	0.115	0.095	0.103	0.373
Insh	57.12	-3.93	2007	10	8	0.092	0.077	0.082	0.498
Kindrum	55.23	-7.71	1997	7	43	0.072	0.058	0.065	0.605
Laggan	56.92	-4.55	2007	6	0	0.115	0.099	0.110	0.328
Langavat	58.04	-6.82	1997	10	81	0.087	0.074	0.078	0.522
Lee	56.90	-2.95	1997	9	35	0.108	0.090	0.096	0.413
Loch	56.85	-3.67	2007	10	83	0.037	0.029	0.031	0.808
Lochy (Benthivore)	56.96	-4.92	2009	2	0	0.188	0.124	0.171	-0.045
Lochy (Planktivore)	56.96	-4.92	2009	5	0	0.123	0.104	0.116	0.288
Lubnaig	56.30	-4.30	1997	10	2	0.103	0.078	0.100	0.387

(Continues)

TABLE 1 | (Continued)

Pop ID	Latitude	Longitude	Sampling year	N	Pa	Ho	He	π	FST
Luichart	57.62	-4.78	2009	10	29	0.102	0.086	0.092	0.440
Maree	57.71	-5.53	2008	10	7	0.131	0.117	0.124	0.243
Meadie	58.33	-4.56	2007	5	2	0.101	0.078	0.089	0.457
Mealt	57.61	-6.18	2009	2	0	0.029	0.018	0.027	0.837
Merkland	58.23	-4.73	1997	10	26	0.068	0.052	0.065	0.601
More	58.28	-4.87	1998	4	0	0.121	0.091	0.113	0.309
Morie	57.75	-4.47	2009	3	0	0.123	0.087	0.116	0.279
Nam Brac	58.38	-5.12	2010	3	0	0.085	0.062	0.081	0.503
naSealga (Benthivore)	57.79	-5.31	2015	5	0	0.089	0.071	0.083	0.497
naSealga (Planktivore)	57.79	-5.31	2015	5	0	0.088	0.072	0.084	0.492
Naver	58.30	-4.35	2011	9	27	0.118	0.100	0.107	0.346
Osgaig	58.05	-5.32	2015	9	35	0.086	0.073	0.079	0.517
Padarn	53.13	-4.13	2020	10	65	0.095	0.082	0.087	0.469
Rannoch (Benthivore)	56.70	-4.23	2010	1	0	0.101	0.050	0.101	0.382
Rannoch (Planktivore)	56.70	-4.23	2010	6	5	0.115	0.098	0.109	0.215
Rannoch (Piscivore)	56.70	-4.23	2010	4	0	0.133	0.111	0.128	0.336
Seil	56.32	-5.55	2021	9	77	0.046	0.039	0.042	0.746
Shin	58.12	-4.57	1997	7	1	0.111	0.095	0.104	0.364
Stack	58.34	-4.92	1997	5	0	0.115	0.102	0.118	0.279
Talla	55.48	-3.40	1997	8	0	0.098	0.084	0.091	0.444
Tarff	57.15	-4.61	2012	8	57	0.049	0.038	0.042	0.744
Tay (Benthivore)	56.52	-4.14	1997	5	1	0.127	0.105	0.122	0.257
Tay (Planktivore)	56.52	-4.14	1997	5	1	0.121	0.107	0.123	0.254
Treig	56.81	-4.73	2007	7	0	0.127	0.124	0.135	0.174
Tummel	56.71	-3.95	2007	9	18	0.126	0.105	0.113	0.308
Uaine	57.52	-5.39	1997	10	0	0.067	0.060	0.064	0.611

Note: Location is either given as country (for outgroups) or Hydrometric Area for Scottish populations (numbered and named per the National River Flow Archive 2014). Pa refers to number of private alleles, He to expected heterozygosity, Ho to observed heterozygosity, π to nucleotide diversity, FST to population-specific FST.

2.6 | Identification of SNPs Correlated to Environmental and Climatic Variables

We conducted a Redundancy Analysis (RDA) using the *vegan* v2.6.2 R package to identify SNPs associated with variation in environmental and climatic variables (Table S1); these represent our putatively adaptive SNP dataset to analyse patterns of adaptive differentiation (Capblancq et al. 2018; Oksanen et al. 2022). RDA was conducted at the individual level as small sample sizes would give potentially inaccurate allele frequency values and our filtering allowed for more confidence in called individual SNP genotypes. As such, individuals in the same lochs, including different ecotypes, were given the same environmental variable values. Environmental data for 19 variables was collected from WorldClim using the *raster* R package v3.5-15 using latitude and longitude coordinates (resolution 2.5 arcmin) (Fick

and Hijmans 2017; Hijmans and van Etten 2012). Bathymetric data on lake surface area, mean and max depth, and altitude was collected for each lake from published work (Fenton, Bean, et al. 2024; Maitland and Adams 2018; Murray and Pullar 1910). An additional bathymetric variable, littoral zone percentage (defined as the area of the lake shallower than 5 m depth), was calculated using available bathymetric maps and *ImageJ v1.50i* (Murray and Pullar 1910; Schneider et al. 2012).

To create a dataset of uncorrelated variables, as collinearity can affect genotype-environment patterns (Ratner 2009), a collinearity matrix was made in R for all environmental and bathymetric variables (Table S2). Variables with correlation coefficient (ρ) of < 0.7 were retained as independent variables and in cases where variables showed high correlation ($\rho > 0.7$) only one variable was retained (Layton et al. 2021) (Table S3).

This resulted in a set of 10 uncorrelated variables (mean diurnal range, isothermality, maximum temperature of the warmest month, mean temperature of the wettest quarter, mean temperature of the driest quarter, annual precipitation, altitude, mean lake depth, lake surface area and the proportion of the lake comprising littoral zone (as %)). Missing data was replaced with the most common genotype at each SNP across all individuals to enable the RDA (Kamvar et al. 2017). An additional partial RDA was run using admixture proportions as a method to control for neutral population structure to ensure our results were not the result of neutral structure mimicking signals of local adaptation. An RDA using population-level allele frequencies was also run to verify that the use of site-level environmental and lake variable data had minimal impact on the end results.

2.7 | Genes and Regions Containing Putatively Adaptive SNPs

To determine if our putative adaptive SNPs were related to functional processes, we investigated the genes and regions containing these SNPs. First, we compared the position of putatively adaptive SNPs with a database of markers for quantitative trait loci (QTL) for various phenotypes from a range of salmonid species mapped to the *Salvelinus* sp. genome (Fenton, Jacobs, et al. 2024). Second, we compared SNP positions to all the annotated genes in the *Salvelinus* sp. genome (ASM291031v2) to identify genes containing putatively adaptive SNPs. Both of these comparisons were carried out using *BEDtools v2.27.1* (Quinlan and Hall 2010), with a limit of the SNP being within ± 100 kbp for QTLs and ± 1 kbp for genes to be considered. We then ran gene ontology term overrepresentation analysis (ORA) for any genes deemed to contain putatively adaptive SNPs using the *TopGO v2.40.0 R package* (Alexa and Rahnenfuhrer 2020) using all genes containing any RAD loci from our ddRADseq data (associated or otherwise) as the comparison dataset. Results were summarised using *REVIGO* (Supek et al. 2011).

2.8 | Analysis of Population Structuring

For analysis of population structuring, we removed any non-neutral SNPs to more accurately reflect neutral evolutionary population genomic patterns (Funk et al. 2012). These were the putatively adaptive SNPs (1104 loci) associated with environmental variables and SNPs in the top 5% of global F_{ST} scores, inferred from *hierfstat* (Goudet 2005). This left a dataset of 22,763 SNPs. Admixture modelling was run using *PopCluster* for $K=2$ to $K=64$ with medium scaling and 10 replicates per K (Wang 2022). The FSTIS and DLK2 estimators were used to infer the optimum number of K (Figure S1). Principal component analyses were run for all SNPs and with just putatively adaptive SNPs using the *adeigenet R package*, with any missing data replaced with the mean allele frequency for that SNP across all individuals.

Distance-based phylogenetic relationships between populations were constructed using an unrooted neighbour-joining (NJ) tree in the *poppr R package v2.9.3* (Kamvar et al. 2015). Any SNPs

with missing data were removed to prevent an effect on the phylogenetic trees. This left a dataset of 5949 SNPs for phylogenetic analyses. The tree was run with 1000 bootstraps with no root. AMOVA was run using the *poppr R package* with a strata list indicating grouping in the N-J tree, Hydrometric Area of origin and population and was run with 1000 repeats.

2.9 | Calculating Genetic Offset

Future forecast climatic data was derived from the Coupled Model Intercomparison Project Phase 5 (CMIP5) under the Representative Concentration Pathways (RCP) scenarios RCP 4.5 and RCP 8.5. The RCP 4.5 scenario assumes CO_2 emissions begin to decline by 2045 while RCP 8.5 assumes that CO_2 levels will continue to rise throughout the 21st century. SNPs associated with climatic variables were determined using an RDA as described earlier. We used climate-associated SNPs to ensure genetic offset did not simply reflect genetic drift rather than adaptive signals (Láruson et al. 2022). To ensure gradientforest ran without errors, we removed any SNPs that showed low variation in allele frequency values (5 or fewer different minor allele frequency values across all populations), leaving us with a dataset of 235 SNPs. Genetic offset was not calculated for Lough Finn and Coniston Water due to an insufficient number (< 2) of individuals. The analysis was also run only including populations having 5 or more individuals to determine the effect of small sample sizes on this analysis (Aguirre-Liguori et al. 2023). To model current genotype-environment relationships, *gradientforest R package v0.1.32* was used with 250 trees (Ellis et al. 2012). This model was then used to transform current and future climatic data based on their importance in explaining genomic variation with the Euclidean distance between these current and future values representing genetic offset. Due to the way *gradientforest* models genotype-environment relationships, a single genetic offset score was calculated for each lake, meaning that there was a single score for lakes containing multiple ecotypes rather than one per ecotype.

2.10 | Lake Sensitivity Metric

Lake sensitivity scores for each lake in Scotland were taken from Maitland and Adams (2018). To summarise, sensitivity scores were calculated based on three bathymetric and location variables: maximum depth of the lake, surface area of the lake, and lake altitude (Winfield et al. 2010). Deeper lakes were deemed to offer more refuge from high surface water temperatures, so were given a lower score. Similarly, higher altitude was deemed to provide some protection from thermal stress, so higher altitude was given a lower score. Small lakes will support smaller populations than larger ones and are impacted more by extreme climate events as a result (Fischer and Lindenmayer 2007), so lakes with a smaller surface area were given a higher score. For maximum lake depth, populations were given the following scores of 1–4 reflecting deep to shallow: maximum depth < 10 m = 4, 10–30 m = 3, 30–100 m = 2, > 100 m = 1. For surface area: < 10 ha = 3, 10–30 ha = 2, > 30 ha = 1. For altitude we categorised as < 10 m = 4, 10–50 m = 3, 50–200 m = 2, > 200 m = 1. Scores for each variable were combined to give a composite score ranging from 3 to 11, with a score of 3 being a lake of

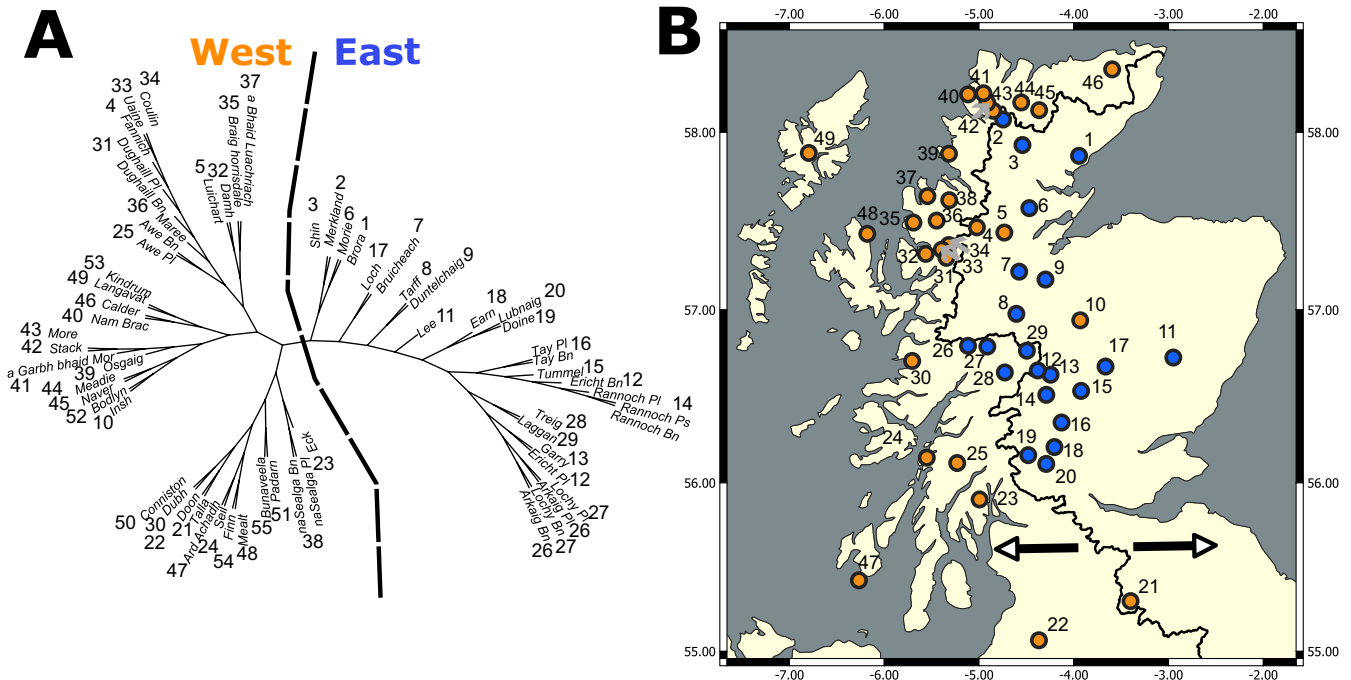


FIGURE 2 | (A) Neighbour-joining plot. NJ plot is split into largely east and west-flowing river system groups and is indicated by the dotted line. (B) shows the geographical distribution of this divide with the black line showing where the drainage divide occurs, populations to the left-hand side appear in west draining river catchments and those to the right in east draining. Populations are coloured by which side of the neighbour-joining tree they appear in, orange for west and blue for east. All outgroup populations are found in the west-flowing group.

least concern for population loss while 11 is of highest concern (Table S4).

3 | Results

3.1 | Summary Statistics

Across the genomic dataset ($N = 24,878$ SNPs) from 410 individuals from 64 populations of Arctic charr, observed heterozygosity for our populations ranged from 0.029 to 0.188 (Figure S2). The Loch Lochy benthivore ecotype population showed the highest observed heterozygosity ($H_o = 0.188$) while the Loch Mealt population had the lowest heterozygosity ($H_o = 0.029$) (Table 1). The number of private alleles ranged from 0 to 164. The Loch Ard Achadh ($P_A = 164$), Lough Bunaveela ($P_A = 127$) and Loch Loch populations ($P_A = 83$) had the highest numbers of private alleles. The number of private alleles negatively correlated with observed heterozygosity ($p < 0.001$, $R^2 = 0.329$).

3.2 | Population Structuring

A neighbour-joining tree broadly separated populations by drainage divide in Scotland (Figure 2). Nearly all populations in river systems that flow to the east coast of Scotland appear in one cluster, while populations in west-flowing river systems appear in a second. The west-flowing system cluster formed two major subgroups. All our outgroup populations from elsewhere in Britain and Ireland appear amongst the west-flowing system population cluster. Within these larger clusters, populations in

the same or nearby river catchments/Hydrometric Areas were usually sister taxa, but some sister taxa were also geographically distant populations such as lochs Eck and na Sealga.

A principal component analysis did not separate major groups of populations; instead, each PC axis only separated out one or a small number of populations and explained a small amount of the total variance. Individuals from lochs Coulin, Uaine, Fannich, Maree and the Dughaill planktivore populations in the northwest of mainland Scotland separate from all other populations along Eigenvector 1 (EV1), which explained 3.62% of the total variance (Figure 3A), while EV2 (explaining 2.57% of the total variance) splits the Loch Loch population, which is in central eastern Scotland, from all others, suggesting high levels of population distinctiveness.

Pairwise genetic differentiation was high between most populations, with a mean pairwise F_{ST} of 0.256 across all populations (Figure 4, Table S5). The lowest genetic differentiation F_{ST} ($F_{ST} = 0.045$) was between Loch Doon and the Talla Reservoir population (which is a conservation refuge population derived from Loch Doon ancestors in the late 1980s (Maitland et al. 2007)). The highest pairwise differentiation was between Lough Finn in Ireland and the Loch Rannoch benthivore population ($F_{ST} = 0.663$); this is probably an artifact of the small number of individuals we had for each population. Several populations such as Loch Loch (mean pairwise $F_{ST} = 0.384$), Loch Seil (mean pairwise $F_{ST} = 0.371$) and Loch Tarff (mean pairwise $F_{ST} = 0.357$) showed particularly high pairwise F_{ST} in all comparisons (all pairwise $F_{ST} > 0.2$). Indeed, comparisons using population-specific F_{ST} showed these populations to be very high as well ($F_{ST} > 0.7$) (Table 1) (Kitada et al. 2021).

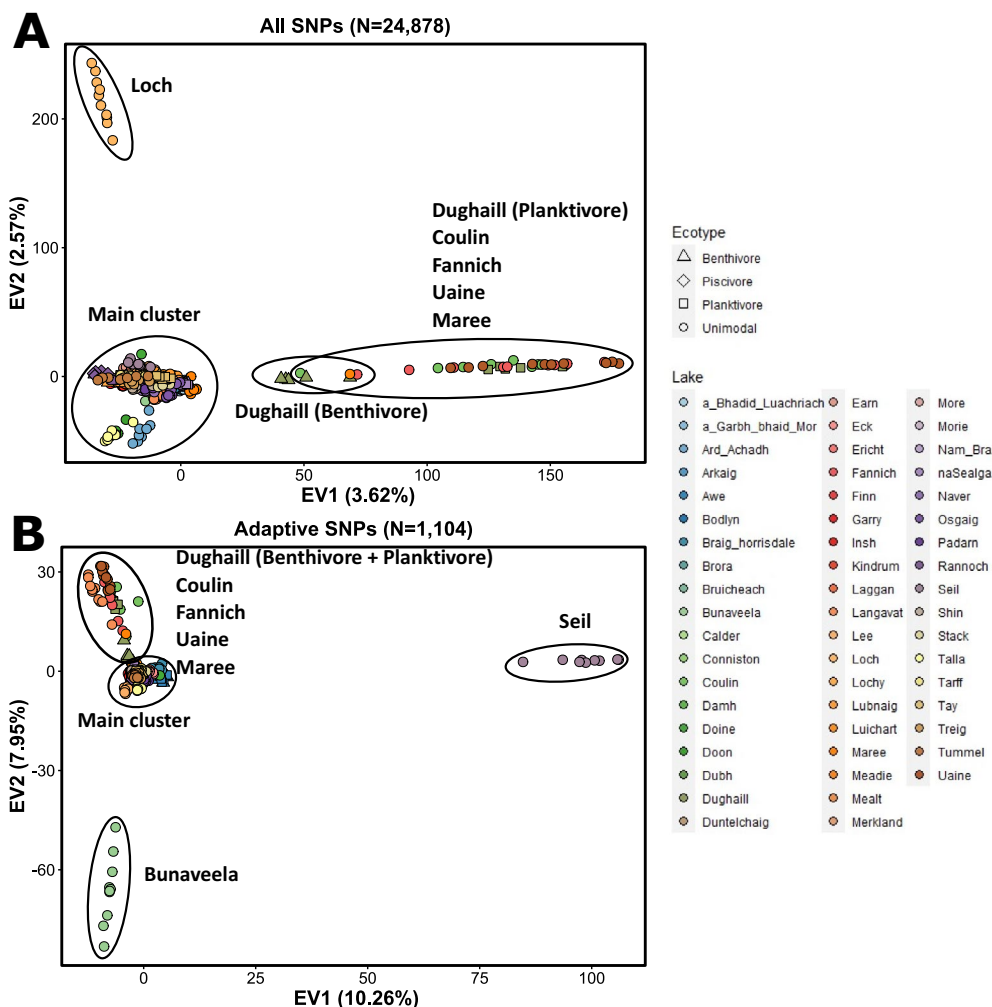


FIGURE 3 | Principal components analyses based on all SNPs (A) ($N = 24,878$) and putatively adaptive SNPs ($N = 1,104$) (B). Individual points are coloured by lake of origin indicated in the legend. Shape is based on population ecotype, triangle for benthivores, diamond for piscivores and squares for planktivores. Unimodal populations are given circles. EV refers to eigenvector.

Admixture analysis was used to infer the number of genetic clusters (K) in the dataset. Fifty distinct clusters were identified as the most likely number in both DLK2 and FSTIS (Figure S1), with most of these clusters consisting of all individuals from a single population and evidence of admixture being rare (Figure 5). An analysis to partition the molecular variance found that most variation was explained by populations within Hydrometric Areas (35.4%). The east-west split found in the topology of the neighbour-joining tree explains <1% of the total variance in the AMOVA model. Hydrometric Area delineations within the east-west split explain 13.6% of the total variance (Table S6).

3.3 | Patterns of Putatively Adaptive Variation

We used a redundancy analysis to identify SNPs associated with changes in climate, altitude and lake bathymetry variables. Approximately 10% of the total variance in the whole dataset was explained by these environmental predictors. Bio3 (Isothermality (BIO2/BIO7)), Bio8 (Mean Temperature of Wettest Quarter) and Bio9 (Mean Temperature of Driest Quarter) all counterposed in the opposite direction of Bio12 (Annual Precipitation) and Bio2

(mean diurnal range) on RDA1 (Figure S3). Altitude and littoral zone percentage counterposed in the opposite direction of mean lake depth and surface on RDA2.

Our analysis showed that all 10 constrained axes were significant in explaining variance and so we then identified SNPs with significant associations across all of them (Table S7). Overall, we identified a set of 1104 SNPs as putatively adaptive to broad scale environmental conditions and lake bathymetry (Table S8). Surface area was the strongest predictor for the highest number of SNPs ($N = 234$) followed by Bio9 (Mean Temperature of Driest Quarter) ($N = 166$) (Table S9). This is reflected in the hierarchy of the importance of environmental variables in explaining adaptive variance with surface area the most important variable followed by Bio9 (Figure S4). When identifying the positions of the putatively adaptive SNPs to annotated genes in the *Salvelinus* sp. genome, we found that 586 known genes contained, or were proximal to, at least one of the 1104 putatively adaptive SNPs (Table S10). GO term analysis showed that amongst the known functions of these genes, a number of processes relating to formation, development and morphogenesis such as skeletal system development (GO:0001501) and sagittal suture morphogenesis (GO:0060367) showed overrepresentation (Table S11).

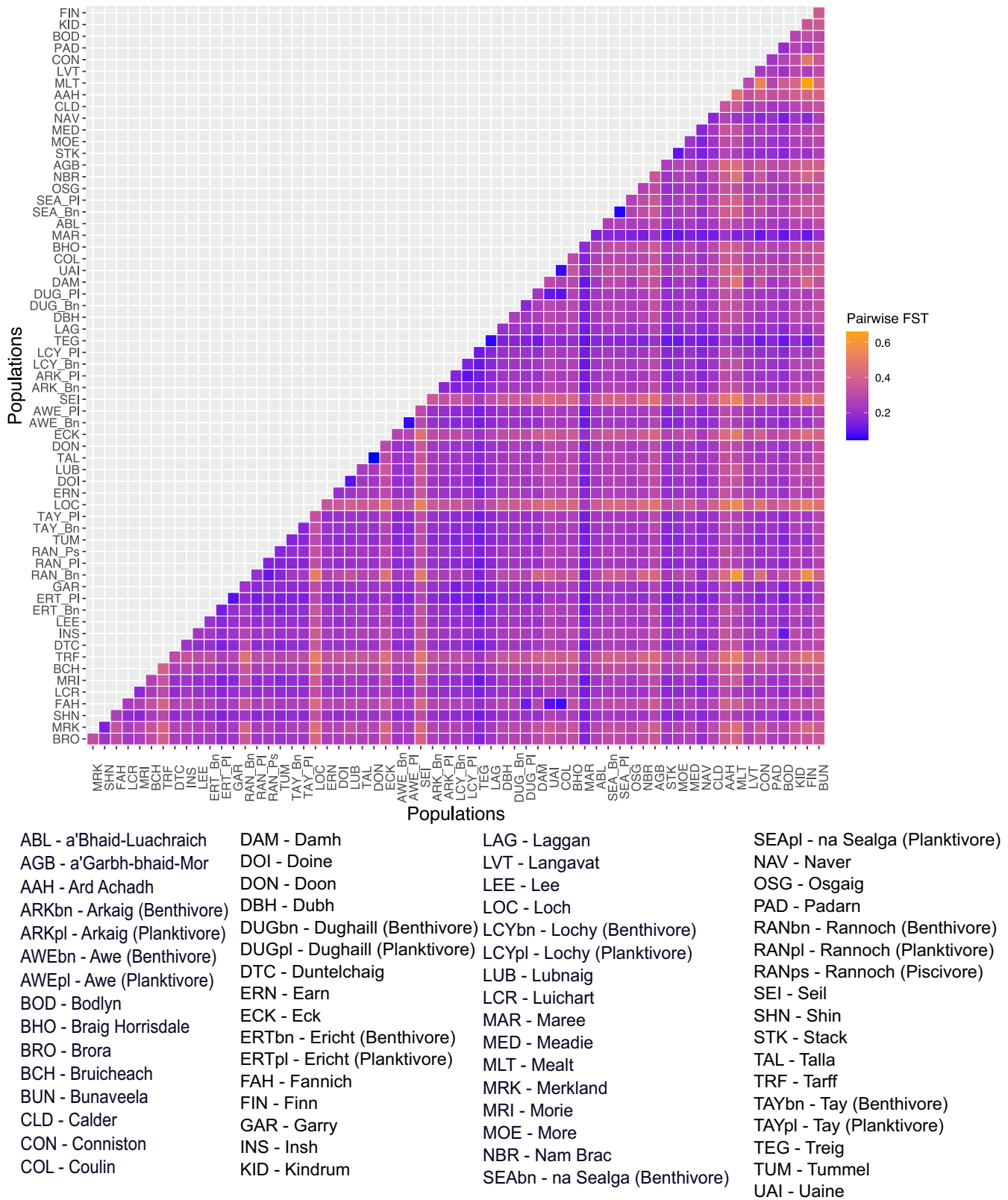


FIGURE 4 | Plot of pairwise F_{ST} comparisons. F_{ST} scores are colours in a range of blue to orange going from low to high. Populations are order in ascending Hydrometric area of origin, order as indicated in Figure 1.

When we compared positions to a database of the known genomic location of salmonid quantitative traits (Fenton, Jacobs, et al. 2024), we found that 35 QTLs contained, or were found close to, those loci we identified as putatively adaptive SNPs

(Table S12). Seven of these QTLs were originally identified from genetic mapping experiments from Arctic charr, while 10 were from lake whitefish (*Coregonus clupeaformis*), 13 from coho salmon (*Oncorhynchus kisutch*) and five from lake trout

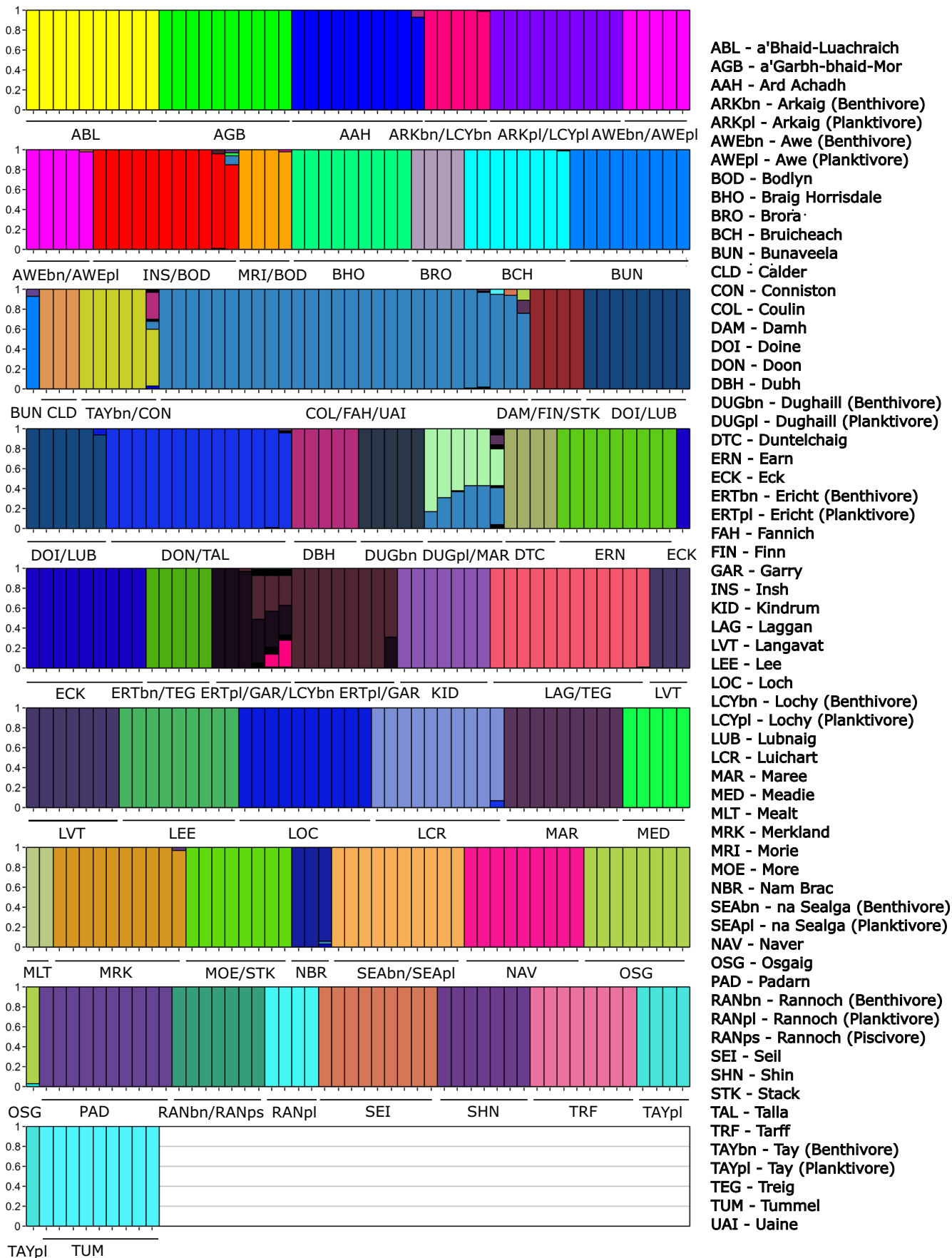


FIGURE 5 | Admixture plot for most likely number of genetic clusters (K), $K=50$. Individuals are ordered by their genetic cluster. Most populations split into their own distinct genetic clusters, with only some proximal populations sharing the same clusters. 3 letter codes for each population are provided. Ecotypes for relevant populations are indicated, bn for benthivore, pl. for planktivore, ps for piscivore.

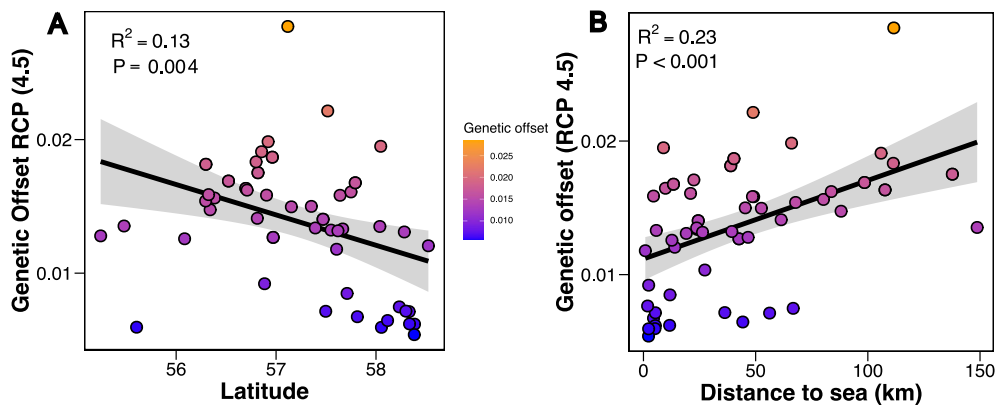


FIGURE 6 | Genetic offset scores under RCP 4.5 scenario for each lake against (A) latitude and (B) distance from the sea in kilometres. Each point, population, is coloured by genetic offset score going from blue to orange as they go from low to high.

(*Salvelinus namaycush*), all closely related species. Over half of the QTLs associated with morphological variation, with seven QTLs related to body length, six related to body shape, three related to body length and a further three related to Fulton's condition factor.

Exploring patterns of adaptive variation through a PCA revealed that EV1 explained 10.3% of the total variance and clearly differentiated the Loch Seil population from all other populations (Figure 3B). EV2 explained 8% of total variance and most clearly separates the Lough Bunaveela population (Ireland) from the other populations in one direction and a group including the lochan Uaine and lochs Coulin, Fannich and Dughail populations in the other direction. Conducting a PCA on the environmental variables, to determine if the patterns identified resulted from distinct environmental conditions, did not show the local conditions for these populations to be outliers in any obvious way (Figure S5). A partial redundancy analysis conducted using admixture proportions to control for neutral structure identified a similar number of total SNPs ($N=1087$), with 818 SNPs being present in the original analysis and the same broad patterns evident across PCs (Figure S6). Similar results were seen on a population-level RDA which identified 1104 SNPs with 743 of the same SNPs identified and the same pattern of single or small groups of populations splitting off on each PC (Figure S7).

3.4 | Susceptibility Metrics

We assessed the potential vulnerability of populations of Arctic charr to a range of threats using three different metrics: genetic offset, lake sensitivity and levels of observed heterozygosity, focusing primarily on the Scottish populations.

We used genetic offset, also known as genomic vulnerability, as a measure of potential maladaptation to changing climate, where higher values indicate elevated predicted maladaptation to future climatic change and risk of extirpation (Fitzpatrick and Keller 2015). A redundancy analysis was used to identify a subset of SNPs ($N=235$) associated with change in climatic variables to be used to generate genetic offset scores. Under the RCP 4.5 emission scenario, genetic offset scores spanned from 0.005 in the Loch Nam Brac population to 0.028 in Loch Insh (Figure 6). We did find a significant relationship between genetic offset

scores and latitude, with southern populations generally showing higher genetic offset scores ($p=0.004$, $R^2=0.13$) (Figure 6A). However, many of the highest offset scores were found in the more northern parts of our distribution (Figure S8). We found a much stronger relationship between offset scores and distance to sea ($p<0.001$, $R^2=0.23$), with populations found in lakes at a further distance from the sea through river systems having higher genetic offset scores (Figure 6B). There was no significant relationship between genetic offset scores and observed heterozygosity ($p=0.249$, $R^2=0.01$) (Figure S9). Similar results were seen when using the more extreme RCP 8.5 scenario to calculate genetic offset scores, for example, genetic offset scores showing a strong correlation with latitude (Figure S10). The same patterns were seen when only calculating genetic offset with populations containing five or more individuals (Figure S11).

We assessed how the local environment affects potential population vulnerability using a metric of lake sensitivity based on lake bathymetry and altitude called lake sensitivity scores from Maitland and Adams (2018). Possible scores ranged from 3 to 11, from least to most vulnerable. Three lochs, lochs Rannoch, Ericht and Treig, had the lowest possible score of 3 in our dataset (Figure 7) whilst the highest score was 8, in lochs Seil and Ard Achadh and the average score was 5. While the different metrics of susceptibility were largely uncorrelated with each other (Table S13), there were some populations such as Loch Seil that performed poorly across all metrics, showing low genetic diversity, a high lake sensitivity score and a high genetic offset score (Figure 7). These populations were not found in any specific geographic region of our dataset and were spread out across the range.

4 | Discussion

Our results highlight how different genetic criteria to delineate priority populations for conservation can lead to differing classification of conservation units. We found that levels of genetic differentiation and genetic distinctiveness were high between almost all populations of Arctic charr in our study. The Hydrometric Area (the catchment or cluster of adjacent catchments) of origin of the population explains a small but meaningful component of the genetic variation which is otherwise very high across all populations. An important finding of our

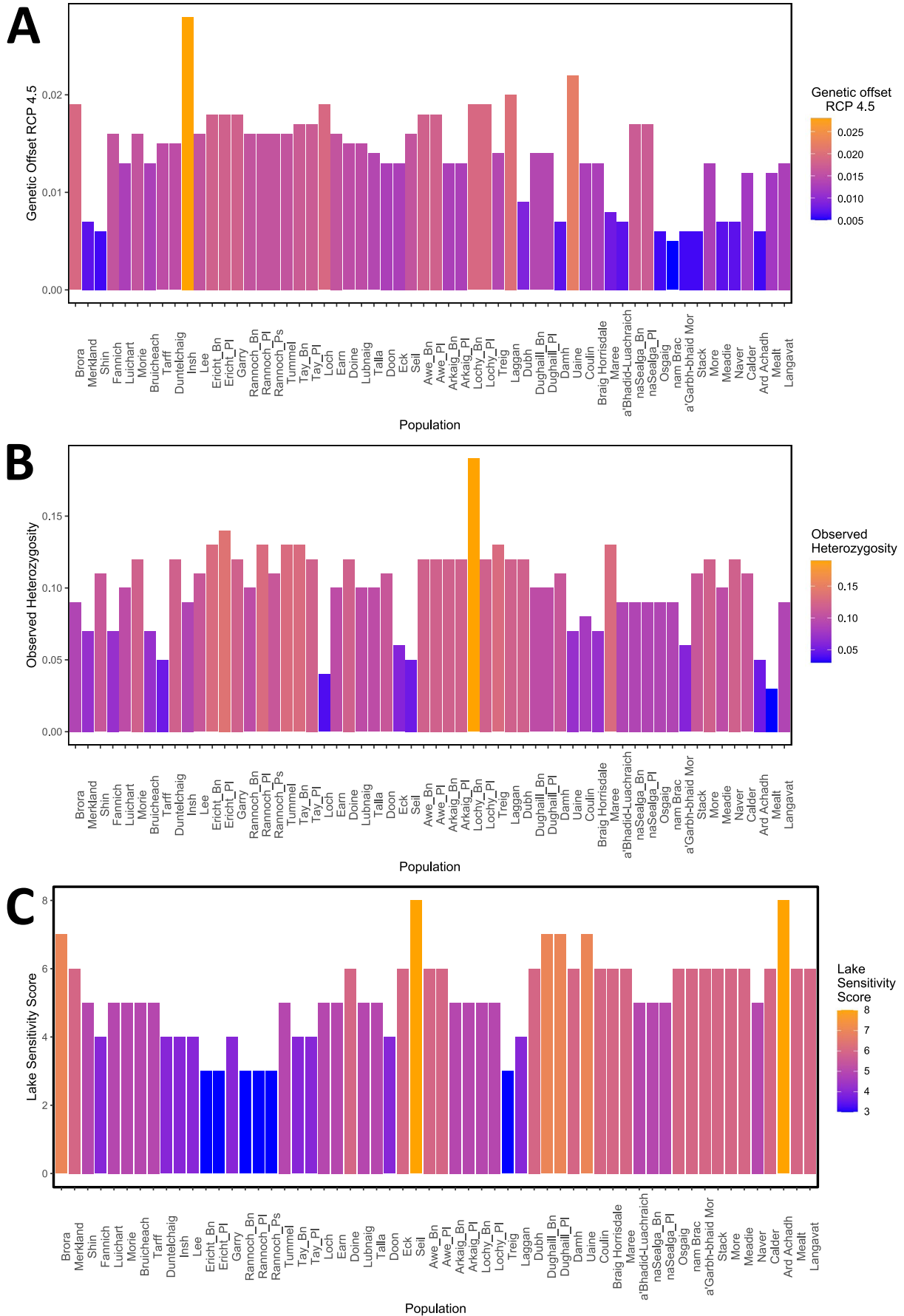


FIGURE 7 | Legend on next page.

FIGURE 7 | Plots showing the susceptibility score for genetic offset (A), heterozygosity (B) and lake sensitivity score (C) for each population in Scotland. Scores are coloured from blue to orange going from low to high values. Due to how each metric is generated, ecotypes in the same lake are given the same genetic offset and lake sensitivity scores. Observed heterozygosity was calculated based on all SNPs ($N=24,878$) while genetic offset was only calculated using SNPs associated with change in climatic variables ($N=235$). Lakes are listed along the x-axis in order of Hydrometric Area. Note that for genetic offset and Lake sensitivity scores, a higher score indicates higher vulnerability while for heterozygosity, a lower score indicates higher vulnerability.

study further identified an important component of putatively adaptive variation that is unique to particular lakes and has the potential to result in differential responses between populations to future selection forces. Given the demonstrated high levels of population differentiation and weak geographic patterns, with distant geographic populations often showing similar levels of differentiation to geographically close ones, we believe geographically large-scale groupings would be inappropriate for conservation unit delineation. We instead focused on prioritising lakes or regions that are conservation-relevant ESUs by exploring and balancing combined information and weighing different ecological and evolutionary components.

4.1 | National-Scale Structuring and Regional Genetic Differentiation

At broader scales, an NJ tree largely clusters populations on either side of the hydrological central divide in Scotland forming east and west groups (Figure 2). While extant Arctic charr in Britain and Ireland are non-anadromous, the species was likely anadromous when it first colonised post-glaciation (Fenton et al. 2023). We expect that these east–west groupings likely represent a historic period of anadromy before the phenotype was lost, with populations in nearby river systems clustering together due to the higher likelihood of contact with one another, a pattern known in freshwater fish species that do not disperse between unconnected river systems (Gómez and Lunt 2007). These groupings, influenced by drainage divide, are similar to that seen in North American populations of brook trout (*Salvelinus fontinalis*) which can be found either side of the Eastern Continent Drainage Divide (Kazyak et al. 2022). However, this east–west grouping only explains a small proportion of the total variance (<1% as inferred from AMOVA) due to high levels of population differentiation and genetic distinctiveness (Table S6).

Analysis of population differentiation showed that most populations were quite differentiated. Many of the pairwise F_{ST} comparisons between populations were notably above the 0.15 value often used to determine considerable differentiation (Table S5) (Frankham et al. 2009). The patterns of high genetic differentiation are reflected in the admixture analysis, with most populations forming their own distinct genetic clusters (Figure 5) and showing little support for clustering of these populations into larger groups. Populations showing lower genetic differentiation and some grouping at a level greater than populations were generally those found in interconnected lakes in the same catchment (e.g., lochs Doine and Lubnaig) or where populations are from the same lake (e.g., the ecotypes at lochs Awe or na Sealga). These findings are consistent with previous studies that found non-anadromous populations of Arctic charr generally show high levels of genetic differentiation and lower levels of genetic

diversity (Salisbury et al. 2023; Shikano et al. 2015). In this, Arctic charr are more structured and differentiated than many other salmonids such as Lake whitefish found in North America (Graham et al. 2022) and Brown trout (*Salmo trutta*) in both England and Ireland (Moccetti et al. 2024; Rodger et al. 2021).

A number of populations in our dataset are highly geographically isolated. Some of these lochs, such as lochs Mealt and Ard Achadh, are on the Hebridean isles of Skye and Islay respectively and therefore isolated from mainland Scotland, while other populations like Lough Bunaveela and Loch Seil are the only populations we have from their Hydrometric Area, and others like Loch Loch are highly separated from other populations in their Hydrometric area. We found that many of these isolated populations, such as Ard Ardchadh, Bunaveela and Loch, tended to also have low heterozygosity with relatively high numbers of private alleles (Table 1). These isolated populations, that we expect have no contemporary gene flow with other populations, have gained unique mutations, and their genetic differentiation may be driven more by genetic drift rather than selective pressures (Funk et al. 2016). The lack of gene flow reflected in the genetic distinctiveness of peripheral populations puts them at particular risk of extirpation due to lower genetic diversity and higher risk of accumulation of deleterious mutations. Whether such isolated populations are of higher conservation value has been debated across species and depends on whether these populations exist in differing environments and their persistence potential (Finn et al. 2009; Lesica and Allendorf 1995); we currently lack this evidence for these populations, and so for some high-impact conservation decisions, more investigation is required.

Several lochs in Scotland have sympatric and ecologically distinct ecotype pairs (Jacobs et al. 2020; Maitland and Adams 2018). We found that levels of differentiation between ecotype pairs were variable (pairwise $F_{ST}=0.05-0.20$). The extent of differentiation is consistent with previously inferred differences in ecotypes' evolutionary histories. Pairs that other research has concluded resulted from admixture after secondary contact of distinct colonising lineages showed higher differentiation (pairwise F_{ST} 0.16–0.20), such as the sympatric ecotype pairs in lochs Tay, Dughail and the Rannoch planktivore versus the Rannoch piscivore/benthivore (Jacobs et al. 2020; Verspoor et al. 2010). In contrast, the ecotypes that have been suggested to be sympatric divergences, particularly lochs Awe and na Sealga, have markedly lower genetic differentiation (pairwise F_{ST} 0.05–0.06) despite clear eco-morphological distinctiveness (Garduño-Paz et al. 2012; Jacobs et al. 2020). While less is known about the lochs Ericht, Lochy and Arkaig ecotype pair origins because these divergences have not been studied in depth before, the levels of genetic differentiation we found here were relatively high (pairwise F_{ST} 0.14–0.17). We speculate that, given this

level of differentiation, these pairs resulted from two post-glacial invasions into the same lake and secondary contact, though this remains to be investigated further (Fenton 2024). Complex and differing evolutionary histories of sympatric ecotypes in postglacial lakes have been recently demonstrated in other areas, such as Arctic charr in Russia and European whitefish in Europe (Aleksyev et al. 2002; Fang et al. 2022; Öhlund et al. 2020) and may be a pervasive feature of contemporary populations even at nearby geographic scales.

4.2 | Putatively Adaptive Differentiation

A potential route to delineating conservation units is to focus on adaptive genes (Hoelzel 2023; Miller et al. 2024) as has been seen in Pacific salmonids in North America where conservation units are based on migration timing which is largely controlled by the *GREB1L/ROCK1* region and is an important phenotype to preserve variability in (Hess et al. 2016; Prince et al. 2017; Waples and Lindley 2018; Waples et al. 2022). With this aim, we identified a set of 1104 SNPs significantly correlated with climatic and lake variables across these populations to explore adaptive potential as well as investigate the genes and regions behind local adaptation across a broad scale. Given that previous studies on Arctic charr have highlighted that adaptive phenotypes, such as head and body shape in divergent ecotype populations, can be highly polygenic (Fenton, Jacobs, et al. 2024; Jacobs et al. 2020), we did not expect to define single regions that could be used for conservation unit delineation. Indeed, we ended up finding a relatively high number of putatively adaptive SNPs and these are distributed across the genome (Table S8).

Over half of these environment-associated SNPs ($N = 624$) were found within or proximal to known genes in the *Salvelinus* sp. genome. We found a high number of terms relating to morphogenesis and development as overrepresented amongst these SNPs. Fully half of the salmonid QTLs that we found containing putatively adaptive SNPs were related directly to morphology (Table S12). Further, 32 of the genes found to contain putatively adaptive SNPs were shared with a recent genome-wide association analysis that identified key genes that underlie head and body shape morphology differences between ecotypes of Arctic charr (Fenton, Jacobs, et al. 2024). While morphology was the most prevalent type of GO term in our dataset, several QTLs identified were related to other important phenotypes such as age of sexual maturation and spawning date, which are known to vary in our populations (Garduño-Paz et al. 2012). This suggests that a number of different aspects of phenotype may be key to local adaptation, although more data is required before concrete conclusions are drawn.

That we identified functional processes and genomic regions associated with head and bone morphology for SNPs associated with temperature variables is consistent with previous studies that indicate that both phenotypes are influenced by temperature across various fish species (Campbell et al. 2021; Hooker et al. 2023; Lema et al. 2019; Riesch et al. 2018). The QTLs containing environment-associated SNPs came from a number of different salmonid species, thereby indicating that key regions underlying important phenotypes are to a meaningful extent shared across species, as has been previously suggested

for salmonids (Fenton, Jacobs, et al. 2024; Jacobs et al. 2017; Salisbury and Ruzzante 2022). Overall, adaptive variation is facilitated through a number of different processes and our results support the role of morphology acting as a key aspect of local adaptation across Arctic charr populations in the British Isles. The functional role of these candidate regions would be a valuable area of future research.

Patterns of differentiation in these putatively adaptive SNPs largely identified distinct single populations rather than larger groupings, particularly Loch Seil and Lough Bunaveela, which cluster separately (Figure 3B). A similar pattern was identified in a similar study on Columbia spotted frogs, where a PCA only separated out single or small groups of populations from the larger distribution (Forester et al. 2022). The distinct populations at putatively adaptive SNPs are not the same as the ones that appear as most distinct in the PCA using all SNPs (Figure 3A). The environmental conditions, lake bathymetry and climatic conditions for these populations were not obviously distinct from the other populations (Figure S5), and therefore the putatively adaptive SNP divergence may be driven by parameters that were unquantified in our study or potentially due to genetic drift.

Our study is the first time many of these populations have been explored genetically, and further evaluation is needed to properly understand the patterns seen. That we identified putatively adaptive SNPs located within genes and regions involved in ecologically relevant phenotypes is a notable finding and has implications for other species with highly isolated populations (Cortázar-Chinarro et al. 2017; Funk et al. 2016). However, caution is warranted on biological, environmental and analytical grounds. For example, it is possible that the putatively adaptive SNPs we identified involved in functional traits are not adaptive and could instead be maladaptive or non-beneficial in contemporary settings (Funk et al. 2012). Environmentally, WorldClim variables are atmospheric measures and it is not demonstrated how accurately they reflect the aquatic conditions of Arctic charr, given variation in lake depth, size and thermal buffering. Finally, high levels of neutral genetic structure can potentially confound gene–environment analysis and mimic signatures of local adaptation (Hay et al. 2022), though here our analysis controlling for neutral structure had broadly similar results (Figure S6). As such, we suggest further evidence is required before putatively adaptive SNPs are used to delineate important conservation decisions such as ESUs.

4.3 | Population Susceptibility

A key part of conservation efforts is identifying populations that may be more vulnerable to loss and are therefore of higher conservation priority. Populations can be more vulnerable for a variety of reasons, from low levels of standing diversity to higher risk due to their environment, with these factors often uncorrelated with one another (Campbell et al. 2021; Fischer and Lindenmayer 2007). Arctic charr is adapted to cold, highly oxygenated environments and as such the species is highly susceptible to climate change, particularly increasing water temperatures and eutrophication (Kelly et al. 2020; Winfield et al. 2008). These risks are thought to be exacerbated at the southern edge of the species range, for example, in North America and central

Europe (Kelly et al. 2020; Layton et al. 2021). However, predicting the potential of population viability and risk is a challenging endeavour impacted by its environment and its population characteristics.

Genetic offset, often referred to as genomic vulnerability, has been suggested to be a metric for potential fitness maladaptation to future climatic change (Fitzpatrick et al. 2021; Layton et al. 2021). Several previous studies on genetic offset in salmonid species such as Arctic charr and Sockeye salmon *Oncorhynchus nerka* in North America have highlighted an important correlation between genetic offset scores and latitude, with more southern populations tending to be at higher vulnerability and therefore needing to undergo a greater magnitude of allelic change to survive in predicted future conditions (Layton et al. 2021; Tigano et al. 2024). Populations in Britain and Ireland are the southern extent of the European range, with the exception of remnant populations in the Alps (Tiberti and Splendiani 2019). Therefore it is perhaps unsurprising that we found more southern populations may be more vulnerable to/require greater allelic change to survive future increase in temperature, which are predicted in the RCIP scenarios (Figure 6a) (Layton et al. 2021). The vulnerability of more southern populations is corroborated by the distribution across Britain, with only a few extant populations found in the southern parts of Scotland and northern England and Wales, and most of the recorded local extinctions occurring in these same regions (Maitland and Adams 2018). However, the highest genetic offset scores we found were in the more northern latitudes, suggesting that genetic offset scores are not simply the product of latitude and the associated differences in climatic conditions and that other factors are contributing.

We found a much stronger relationship between genetic offset scores and geographic distance to the sea, with populations in lakes further to the sea having higher genetic offset scores (Figure 6B). This relationship suggests that more inland populations have had reduced levels of gene flow since colonisation, as seen in other fish species (Gouthier et al. 2023), and has resulted in them requiring a greater degree of genetic change to not be maladapted to future climatic conditions. Notably, genetic offset scores were not driven by levels of observed heterozygosity (with heterozygosity used here as an indirect proxy for the allele frequencies used in the gene–environment analysis [Figure S9]), suggesting scores were largely not driven by standing levels of genetic variation (Láruson et al. 2022; Tigano et al. 2024).

It is worth noting that there are several limitations of our genetic offset analysis due to our dataset (Láruson et al. 2022; Rellstab et al. 2021). Firstly, the small sample sizes for many populations mean there is a risk we did not accurately determine allele frequencies. Secondly, the use of population-level climatic variable values for the initial RDA, which was conducted at the individual level, introduces pseudo-replication and potentially biases the analyses although we found broadly similar results when running the RDA with population-level allele frequencies instead (Figure S7). Further issues of pseudoreplication occur when running the genetic offset analysis as ecotypes are considered different populations but were given the same climatic variable values (as being at the same location). Additionally, as previously mentioned, WorldClim variables are atmospheric

measures and so may not accurately reflect the within-lake environments experienced by fish.

While genetic offset and observed heterozygosity did not correlate in our dataset, observed heterozygosity still represents an important indicator of population vulnerability. Low levels of standing genetic variation suggest a population is less able to adapt to changes in their environment and are at higher risk of reduced population fitness due to inbreeding depression (Reed and Frankham 2003). Similarly, populations in small lake environments may also be of concern for potential loss. Deeper, larger lakes hold more diverse populations both genetically and phenotypically (Fenton, Bean, et al. 2024; Recknagel et al. 2017), while small lakes have lower diversity and provide less refuge from the danger of rising water temperatures (Campbell et al. 2021; Fischer and Lindenmayer 2007).

Overall, our metrics of susceptibility were largely independent of each other, allowing us to investigate vulnerability across multiple axes (Table S13) as is valuable in research focusing on determining conservation priority (Auber et al. 2022; Du et al. 2024). Looking across our metrics did not highlight any specific geographic regions of particular concern; the worst scoring populations in each metric were found distributed across our range (Figure 7). While we did note some populations that performed poorly across all metrics, such as Loch Seil, our analyses suggest that populations need to be assessed individually to determine their viability and vulnerability and there are no broader patterns such as an effect of Hydrometric Area. Our analyses do show, however, that the latitudinal position of a population, its distance from the sea and overall lake size can also be used to give indications of potential vulnerability and can be used in a wider-scale analysis to identify potential regions of higher vulnerability or viability in the species distribution. Further work should look to include other potential threats to the species, such as lake acidification, hydroelectric work or non-native introductions to better assess population vulnerability (Knudsen et al. 2016; Maitland et al. 2007). Rather than trying to define a single definitive statistic to determine vulnerability, the investigation of vulnerability across a range of metrics can help us identify vulnerable populations in ways that may be missed by the use of a single metric.

4.4 | Delineation of ESUs and MUs

Conservation units have had numerous definitions and the specifics used are often reflective of the system it is being applied to (Coates et al. 2018; De Guia and Saitoh 2007; Moritz 1994). Many studies that define ESUs and MUs delineate them in a hierarchical fashion, with larger ESUs containing multiple demographically independent populations that are considered to be MUs (Funk et al. 2012; Galland et al. 2021; Shaney et al. 2020). However, we argue that this dataset generated for the highly diverse Arctic charr is a good example of how the criteria for conservation units should be best applied for the species and system in question. The high levels of population differentiation seen in our admixture, F_{ST} and AMOVA analyses (Figures 4 and 5, Table S6) limit the usefulness of delineating broader ESUs, which would be the major branches in our neighbour-joining tree for example (Figure 2). Given the high level of genetic

TABLE 2 | Conservation unit classifications for populations in our dataset as Evolutionarily Significant Units (ESUs) or management units (MUs).

Population(s)	Conservation unit type	Delineation criteria for ESU
a'Bhaid-Luachraich	MU	
a'Garbh-bhaid Mor	MU	
Ard Achadh	MU	
Arkaig (Benthivore)	ESU	Diverged ecotype pop.
Arkaig (Planktivore)	ESU	Diverged ecotype pop.
Awe (Benthivore)	ESU	Diverged ecotype pop.
Awe (Planktivore)	ESU	Diverged ecotype pop.
Braig horrisdale	MU	
Brora	MU	
Bruicheach	ESU	Only extant HA pop.
Calder	MU	
Coulin, Dughail (Planktivore), Fannich, Uaine	ESU	Diverged ecotype pop.
Damh	MU	
Doine	ESU	Diverged ecotype pop.
Doon, Talla	ESU	Only extant HA pop.
Dubh	MU	
Dughail (Benthivore)	ESU	Diverged ecotype pop.
Duntelchaig	MU	
Earn	ESU	Only extant HA pop.
Eck	ESU	Only extant HA pop.
Ericht (Benthivore)	ESU	Diverged ecotype pop.
Ericht (Planktivore), Garry	ESU	Diverged ecotype pop.
Insh	MU	
Laggan, Treig	MU	
Langavat	MU	
Lee	ESU	Only extant HA pop.

(Continues)

TABLE 2 | (Continued)

Population(s)	Conservation unit type	Delineation criteria for ESU
Loch	MU	
Lochy (Benthivore)	ESU	Diverged ecotype pop.
Lochy (Planktivore)	ESU	Diverged ecotype pop.
Lubnaig	ESU	
Luichart	MU	
Maree	ESU	Diverged ecotype pop.
Meadie	MU	
Mealt	MU	
Merkland	ESU	Diverged ecotype pop.
More	ESU	Diverged ecotype pop.
Morie	MU	
Nam Brac	MU	
naSealga (Benthivore)	ESU	Diverged ecotype pop.
naSealga (Planktivore)	ESU	Diverged ecotype pop.
Naver	MU	
Osgaig	MU	
Rannoch (Benthivore)	ESU	Diverged ecotype pop.
Rannoch (Piscivore)	ESU	Diverged ecotype pop.
Rannoch (Planktivore)	ESU	Diverged ecotype pop.
Seil	MU	
Shin	ESU	Diverged ecotype pop.
Stack	ESU	Diverged ecotype pop.
Tarff	MU	
Tay (Benthivore)	ESU	Diverged ecotype pop.
Tay (Planktivore)	ESU	Diverged ecotype pop.
Tummel	MU	

Note: Delineation criteria for groups as ESUs are listed; in our study this was either due to being a diverging ecotype population or being the only extant population in a Hydrometric Area.

differentiation and important site-specific variation, we suggest a more population-level approach to delineations is most appropriate here; this is also compatible with existing frameworks, such as the SSSIs, which work on a site-based level (Mainstone et al. 2018). Our study aims to use evidence from the conservation genomic analyses described herein to assess Arctic charr populations (including sympatric ecotype populations) for their relative conservation value in Scotland. Our findings feed into both the genetic conservation unit and the place-based perspective, given we examined both the genetic diversity and the role of local environment.

We defined an ESU to be a population or group of populations that showed demographic independence to other groups and show evidence of adaptive differentiation, functional diversity and/or geographic importance (Funk et al. 2012; De Guia and Saitoh 2007). We considered an MU to be populations or groups of populations that show demographic independence to one another (Funk et al. 2012; Moritz 1994). This system is aligned with the rationale used for the delineation of partial and full ESUs that was suggested by De Guia and Saitoh (2007). In that case, partial ESUs show demographic independence but lack evidence of adaptive differentiation; although due to limited take up of the 'partial ESU' terminology, we proceed with the more standardly used MUs, which have the same definition. We consider the long-term conservation value for natural heritage to be higher for ESUs than MUs, as ESUs hold more adaptive and evolutionary potential (Funk et al. 2012; Moritz 1994).

To delineate MUs, we first used the patterns of genetic differentiation to determine the number of demographically independent populations or groups of populations (Allendorf et al. 2022; Palsbøll et al. 2007). Given our findings, we suggest that populations should be considered within the same conservation units (whether that is ESU or MU) when they are in close proximity (< 20 km), show notable gene flow, represent prior conservation translocations (Maitland et al. 2007), or share notable close genetic history, for example, recently diverged ecotypes (Jacobs et al. 2020). Amongst the determined MUs, we then consider the evidence of substantial adaptive differentiation, functional diversity and/or geographic importance that would warrant elevation to ESU status.

To delineate ESUs, we considered multiple lines of evidence, including phenotypic, environmental and geographic information, in addition to genetic data (Waples et al. 2022; Waples and Lindley 2018). We first explored patterns of adaptive differentiation through the identification of non-neutral and environment-associated SNPs, as other studies have done (Forester et al. 2022; Matala et al. 2014; Miller et al. 2024). While the identification of adaptive variation undeniably has importance in the delineation of conservation units (Hoelzel 2023; Waples and Lindley 2018; Waples et al. 2022), as discussed earlier, we caution the use of the SNPs we identified as putatively adaptive for the primary basis of our conservation unit delineations. Second, we deemed ecotype populations, either in multimodal lakes or parapatric divergences, as having high conservation value and should be considered ESUs due to their importance to the functional diversity of the species and its evolutionary future (Jonsson and Jonsson 2001; Maitland and Adams 2018) (Table 2). These include ecotypes that do not show clear genetic differentiation

but have reproductive isolation and ecological distinctiveness. These represent important incipient diverging groups, although not necessarily the same evolutionary potential as those resulting from secondary contact (Adams and Huntingford 2004; Adams et al. 1998; Alexander and Adams 2000; Garduño-Paz et al. 2010; Jacobs 2018; Jonsson and Jonsson 2001; Kettle-White 2001). While some morphological and behavioural variation might be due to phenotypic plasticity (Hooker et al. 2023; Kristjánsson et al. 2018), the precautionary principle motivates us to nonetheless give some conservation priority to those populations. Third, populations that are the only extant population in their respective Hydrometric Area we also considered as ESUs based on geographic importance. These have a higher potential to gain unique and important adaptive variation due to their extreme isolation, and their loss would significantly affect the geographic range of the species (Lesica and Allendorf 1995).

The outcome of this assessment for ESUs delineated 29 groups (Table 2). The ecotype-polymorphic populations we consider to be ESUs are Awe (benthivore and planktivore), Arkaig (benthivore and planktivore), Doine, Dughail (benthivore and planktivore), Erich (benthivore and planktivore), Lochy (benthivore and planktivore), Lubnaig, Maree, Merkland, More, na Sealga (benthivore and planktivore), Shin, Stack, Rannoch (benthivore, planktivore and piscivore) and Tay (benthivore and planktivore). The loch Dughail planktivore ESU also contains the lochs Coulin and Fannich and lochan Uaine populations, due to patterns of genetic similarity seen in admixture and PCA analyses influenced by a shared ancestral history (Fenton 2024). The known ecotypes at lochs Maree and Stack (Adams et al. 2008) should still be considered separate ESUs; though they were not directly studied in this work, they should be investigated in future. ESUs for Arctic charr lakes that are the only populations in their respective Hydrometric Areas are in lochs Eck, Doon, Bruicheach, Lee and Earn. We suggest the Loch of Girsta, which is the sole extant population on the Shetland Islands, represents an ESU on the same criteria but as it was not included in our dataset it needs further investigation (Maitland and Adams 2018).

We define all remaining groups with the working designations of MUs. In the scope of our study, this includes 24 populations that lie across the extent of Scotland (Figure 1, Table 2).

These designations we propose are well informed by genomic data and contemporary analyses but should be considered recommendations about priority and not considered definitive. More data on these populations should be used to further our understanding of the contemporary variation and relationship between populations, both within our dataset and those not currently covered, to help delineate conservation units. There are many other charr populations in Scotland that are not in our dataset and would benefit from genetic characterising; Maitland and Adams (2018) lists 187 lochs with extant or likely extant populations of Arctic charr in Scotland. Other factors such as unique historical lineages (Jacobsen et al. 2022; Moore et al. 2015) and adaptively important phenotypes, for example, a difference in spawning (Walker 2007), foraging/diet (Garduño-Paz and Adams 2010) or habitat usage (Kristjánsson et al. 2012) would also be considered evolutionarily significant (Zhao et al. 2020). Given that we now have an understanding of

national-scale patterns of genetic differentiation and structuring, and thus we have suggested how conservation units can be delineated, more focus should go into determining conservation priority from a management and policy perspective.

The focus on protecting environments and landscapes (Mainstone et al. 2018) already gives some protection to some Arctic charr populations in Scotland, irrespective of evolutionary genetics. For example, lochs Eck and Doon are included as notified features within protected sites (such as Sites of Special Scientific Interest) and are therefore protected in law (Bean et al. 2018; Mainstone et al. 2018). Lochs Doon, Lee and Girlsta have also been flagged previously as sites that may be of higher risk of loss, due to factors such as lake acidification (Maitland et al. 1991, 2007). A number of lochs, such as Maree and Rannoch, are located within sites classified as Special Areas of Conservation under the EU Habitats Directive and Special Protection Areas (under the EU Birds Directive) to protect other habitats and species present, which also benefits Arctic charr (McLeod et al. 2009). However, it is evident that current landscape-based protection is not encompassing the ESUs and MUs defined here. Given the high diversity seen in Arctic charr, we urge that many of these ESU sites we have defined be considered for special protection, for example, as SSSIs. While the formal designation of ESU population as a notified feature based purely on our findings is not always likely, including some ESU populations to the list of notified features in existing SSSIs, already protected for other purposes, is certainly appropriate. This is the case for the populations in lochs Maree and Calder, for example. Further, protection of supporting habitats through Water Framework Directive (WFD) River Basin Management Plans, which focus on water regulation and planning activities, and the application of clear regulation to prevent the introduction of non-native species can all contribute to the health and security of existing populations (SEPA 2021).

Our results strongly urge against the movement of Arctic charr between lakes in different river catchments, particularly those on the opposite sides of the east–west drainage divide. This aligns with previous suggestions for the conservation of the species (Maitland et al. 2007), due to the lack of contemporary connection between populations. In other Scottish salmonids of conservation concern, such as European whitefish (powan), translocations have been used with good success (Adams et al. 2014; Crotti et al. 2021; Lyle et al. 2006). If translocations are applied in Arctic charr, then we suggest that population genetic patterns should guide the practice with regard to population profiles. At southern portions of the species range in the UK, where populations are of high concern due to rapid declines, translocations are considered as a possible feature of conservation strategy and have already been applied to Loch Doon (Lavictoire et al. 2025; Maitland et al. 2007). The general high levels of genetic differentiation suggest that the conservation of most populations needs to be handled independent of one another.

5 | Conclusions

Populations clustered into their river catchments and on a wider scale by river system flow; however, levels of genetic

differentiation were high, even between populations in the same river systems, suggesting low levels or a lack of gene flow and that most populations are demographically independent from one another. Identification of putatively adaptive SNPs highlighted a number of genes and regions with roles in local adaptation that warrant further investigation to determine the importance of their role. Our metrics of population vulnerability highlighted the value of using multiple metrics, demonstrating how the geographical location and size of lake environment can affect population vulnerability. We focused conservation efforts on populations showing important phenotypes to preserve, such as ecological versatility, as well as those that show important adaptive potential. Our study highlights the importance of wide-scale datasets when delineating conservation units, as well as the challenges for appropriate delineations in highly diverse species.

Acknowledgements

The authors thank Eric Verspoor and Ron Greer for their help in collecting and providing samples for 12 of the Scottish lochs included in this dataset; and Peter Koene, Hannele Honkanen, Rowan Smith, Jamie Carruth, Sandy Macfarlane and Ruaidhri Forrester for assistance with fieldwork. We thank David Wood and the RSPB Oa reserve team for facilitating access to Loch Ard Achadh, and Alastair Stephen for support with the applied project development. Thanks to Maria Capstick for teaching the lab protocols used in this project and Kara Layton for her help and insight with the genetic offset analysis. Thanks also to Arne Jacobs for providing existing data for eight of the studied lakes and for support with bioinformatic processing and analysis. This research was supported by NatureScot, SSE, and the School of Biodiversity, One Health & Veterinary Medicine at the University of Glasgow.

Funding

This work was supported by NatureScot, University of Glasgow, and Scottish and Southern Energy (SSE).

Conflicts of Interest

The authors declare no conflicts of interest.

Data Availability Statement

Data accessibility: Relevant sample meta-data, environmental information, SNP VCF files and code used in the analyses are archived on the University of Glasgow's permanent data repository, Enlighten (<https://doi.org/10.5525/gla.researchdata.2114>). Demultiplexed ddRADseq data are available on NCBI Short Read Archive (SRA) (BioProjects PRJNA1061680, PRJNA607173). *Benefits Generated:* A research collaboration was developed with scientists and fisheries/conservation managers and stakeholders. All collaborators who provided material and feedback and/or collection support are included as co-authors or in acknowledgements. The results of research are shared publicly, with the relevant conservation communities, and the broader scientific community. This research addresses a priority concern for Scotland, in this case the conservation of organisms being studied. Benefits from this research accrue from the sharing of our data and results on public databases as described above.

References

Adams, C. E., C. W. Bean, D. Fraser, and P. S. Maitland. 2007. "Conservation and Management of the Arctic Charr: A Forward View." *Ecology of Freshwater Fish* 16, no. 1: 2–5. <https://doi.org/10.1111/j.1600-0633.2006.00180.x>.

- Adams, C. E., D. Fraser, F. A. Huntingford, R. B. Greer, C. M. Askew, and A. F. Walker. 1998. "Trophic Polymorphism Among Arctic Charr From Loch Rannoch, Scotland." *Journal of Fish Biology* 52, no. 6: 1259–1271. <https://doi.org/10.1006/jfbi.1998.0676>.
- Adams, C. E., D. J. Hamilton, I. McCarthy, et al. 2006. "Does Breeding Site Fidelity Drive Phenotypic and Genetic Sub-Structuring of a Population of Arctic Charr?" *Evolutionary Ecology* 20, no. 1: 11–26. <https://doi.org/10.1007/s10682-005-2489-4>.
- Adams, C. E., and F. A. Huntingford. 2004. "Incipient Speciation Driven by Phenotypic Plasticity? Evidence From Sympatric Populations of Arctic Charr." *Biological Journal of the Linnean Society* 81, no. 4: 611–618. <https://doi.org/10.1111/j.1095-8312.2004.00314.x>.
- Adams, C. E., A. A. Lyle, J. A. Dodd, et al. 2014. "Translocation as a Conservation Tool: Case Studies From Rare Freshwater Fishes in Scotland." *Glasgow Naturalist* 26: 17–24.
- Adams, C. E., and P. S. Maitland. 2007. "Arctic Charr in Britain and Ireland – 15 Species or One?" *Ecology of Freshwater Fish* 16, no. 1: 20–28. <https://doi.org/10.1111/j.1600-0633.2006.00179.x>.
- Adams, C. E., A. J. Wilson, and M. M. Ferguson. 2008. "Parallel Divergence of Sympatric Genetic and Body Size Forms of Arctic Charr, *Salvelinus alpinus*, From Two Scottish Lakes." *Biological Journal of the Linnean Society* 95, no. 4: 748–757. <https://doi.org/10.1111/j.1095-8312.2008.01066.x>.
- Aguirre-Liguori, J. A., A. Morales-Cruz, B. S. Gaut, and S. Ramírez-Barahona. 2023. "Sampling Effect in Predicting the Evolutionary Response of Populations to Climate Change." *Molecular Ecology Resources* 25, no. 5: e13828. <https://doi.org/10.1111/1755-0998.13828>.
- Alekseyev, S. S., V. P. Samusenok, A. N. Matveev, and Y. M. Pichugin. 2002. "Diversification, Sympatric Speciation, and Trophic Polymorphism of Arctic Charr, *Salvelinus alpinus* Complex, in Transbaikalia." *Environmental Biology of Fishes* 64: 97–114. <https://doi.org/10.1023/A:1016050018875>.
- Alexa, A., and J. Rahnenfuhrer. 2020. "topGO: Enrichment Analysis for Gene Ontology (v2.40.0)." <https://bioc.ism.ac.jp/packages/3.13/bioc/manuals/topGO/man/topGO.pdf>.
- Alexander, G. D., and C. E. Adams. 2000. "The Phenotypic Diversity of Arctic Charr, *Salvelinus alpinus*, (Salmonidae) in Scotland and Ireland." *Aqua, International Journal of Ichthyology* 4, no. 2: 77–88.
- Allendorf, F. W., W. C. Funk, S. N. Aitken, M. Byrne, and G. Luikart. 2022. *Conservation and the Genetics of Populations*. 3rd ed. Blackwell.
- Auber, A., C. Waldock, A. Maire, et al. 2022. "A Functional Vulnerability Framework for Biodiversity Conservation." *Nature Communications* 13, no. 1: 4774. <https://doi.org/10.1038/s41467-022-32331-y>.
- Barthelemy, N., R. Hynes, G. Hutchinson, and P. A. Prodöhl. 2023. "Taxonomy and Phylogeography of the Irish Arctic Char (*Salvelinus alpinus*)." www.epa.ie.
- Bean, C. W., C. P. Mainstone, R. A. Hall, T. W. Hatton-Ellis, A. S. Lee, and P. J. Boon. 2018. "Guidelines for the Selection of Biological SSSIs. Part 2: Detailed Guidelines for Habitats and Species Groups. Chapter 19 Freshwater Fish." In *Guidelines for the Selection of Biological SSSIs*, vol. 2. Joint Nature Conservation Committee.
- Blain, S. A., C. E. Adams, P. Amundsen, R. Knudsen, and L. Chavarie. 2025. "Which Ecological Factors Influence the Level of Intraspecific Diversity Within Post-Glacial Fishes? A Case Study Using *Coregonus* and *Salvelinus*." *Freshwater Biology* 70, no. 1: e14357. <https://doi.org/10.1111/fwb.14357>.
- Bolger, A. M., M. Lohse, and B. Usadel. 2014. "Trimmomatic: A Flexible Trimmer for Illumina Sequence Data." *Bioinformatics* 30, no. 15: 2114–2120. <https://doi.org/10.1093/bioinformatics/btu170>.
- BRIG. 2007. "Report on the Species and Habitat Review (Report by the Biodiversity Reporting and Information Group (BRIG) to the UK Standing Committee)."
- Campbell, C. S., C. E. Adams, C. W. Bean, and K. J. Parsons. 2017. "Conservation Evo-Devo: Preserving Biodiversity by Understanding Its Origins." *Trends in Ecology & Evolution* 32, no. 10: 746–759. <https://doi.org/10.1016/j.tree.2017.07.002>.
- Campbell, C. S., C. E. Adams, C. W. Bean, N. Pilakouta, and K. J. Parsons. 2021. "Evolvability Under Climate Change: Bone Development and Shape Plasticity Are Heritable and Correspond With Performance in Arctic Charr (*Salvelinus alpinus*)." *Evolution and Development* 23, no. 4: 333–350. <https://doi.org/10.1111/ede.12379>.
- Capblancq, T., K. Luu, M. G. B. Blum, and E. Bazin. 2018. "Evaluation of Redundancy Analysis to Identify Signatures of Local Adaptation." *Molecular Ecology Resources* 18, no. 6: 1223–1233. <https://doi.org/10.1111/1755-0998.12906>.
- Casacci, L. P., F. Barbero, and E. Balletto. 2014. "The 'Evolutionarily Significant Unit' Concept and Its Applicability in Biological Conservation." *Italian Journal of Zoology* 81, no. 2: 182–193. <https://doi.org/10.1080/11250003.2013.870240>.
- Cavers, S., J.-P. Lobo-Guerrero, and A. Perry. 2022. "Pilot Study to Find Potential New Sites for Genetic Conservation Units for Native Trees in Scotland." NatureScot Research Report, 1306.
- CEN. 2015. "Water Quality—Sampling of Fish With Multi-Mesh Gillnets." ICS.
- Coates, D. J., M. Byrne, and C. Moritz. 2018. "Genetic Diversity and Conservation Units: Dealing With the Species-Population Continuum in the Age of Genomics." *Frontiers in Ecology and Evolution* 6: 165. <https://doi.org/10.3389/fevo.2018.00165>.
- Cortázar-Chinarro, M., E. Z. Lattenkamp, Y. Meyer-Lucht, E. Luquet, A. Laurila, and J. Höglund. 2017. "Drift, Selection, or Migration? Processes Affecting Genetic Differentiation and Variation Along a Latitudinal Gradient in an Amphibian." *BMC Evolutionary Biology* 17, no. 1: 189. <https://doi.org/10.1186/s12862-017-1022-z>.
- Crotti, M., C. E. Adams, E. C. Etheridge, et al. 2020. "Geographic Hierarchical Population Genetic Structuring in British European Whitefish (*Coregonus lavaretus*) and Its Implications for Conservation." *Conservation Genetics* 21, no. 5: 927–939. <https://doi.org/10.1007/s10592-020-01298-y>.
- Crotti, M., E. Yohannes, I. J. Winfield, A. A. Lyle, C. E. Adams, and K. R. Elmer. 2021. "Rapid Adaptation Through Genomic and Epigenomic Responses Following Translocations in an Endangered Salmonid." *Evolutionary Applications* 14, no. 10: 2470–2489. <https://doi.org/10.1111/eva.13267>.
- Danecek, P., A. Auton, G. Abecasis, et al. 2011. "The Variant Call Format and VCFtools." *Bioinformatics* 27, no. 15: 2156–2158. <https://doi.org/10.1093/bioinformatics/btr330>.
- Danecek, P., J. K. Bonfield, J. Liddle, et al. 2021. "Twelve Years of SAMtools and BCftools." *GigaScience* 10, no. 2: giab008. <https://doi.org/10.1093/gigascience/giab008>.
- de Apolônio Silva Oliveira, D., W. Decraemer, T. Moens, G. A. P. dos Santos, and S. Derycke. 2017. "Low Genetic but High Morphological Variation Over More Than 1000 Km Coastline Refutes Omnipresence of Cryptic Diversity in Marine Nematodes." *BMC Evolutionary Biology* 17, no. 1: 71. <https://doi.org/10.1186/s12862-017-0908-0>.
- De Guia, A. P. O., and T. Saitoh. 2007. "The Gap Between the Concept and Definitions in the Evolutionarily Significant Unit: The Need to Integrate Neutral Genetic Variation and Adaptive Variation." *Ecological Research* 22, no. 4: 604–612. <https://doi.org/10.1007/s1128-4-006-0059-z>.
- Denys, G. P. J. 2021. "A Comment on "Morphologic and Genetic Characterization of Corsican and Sardinian Trout With Comments on *Salmo* Taxonomy" by Delling et al. (2020): Protected Tyrrhenian Trouts Must Be Named." *Knowledge and Management of Aquatic Ecosystems* 422: 6. <https://doi.org/10.1051/kmae/2021006>.

- Department of Agriculture, Environment and Rural Affairs (DAERA), Scottish Government, Welsh Government, & UK Government. 2025. "Blueprint for Halting and Reversing Biodiversity Loss: The UK's National Biodiversity Strategy and Action Plan for 2030."
- Des Roches, S., D. M. Post, N. E. Turley, et al. 2018. "The Ecological Importance of Intraspecific Variation." *Nature Ecology & Evolution* 2, no. 1: 57–64. <https://doi.org/10.1038/s41559-017-0402-5>.
- Doenz, C. J., A. K. Krähenbühl, J. Walker, O. Seehausen, and J. Brodersen. 2019. "Ecological Opportunity Shapes a Large Arctic Charr Species Radiation." *Proceedings of the Royal Society B: Biological Sciences* 286, no. 1913: 20191992. <https://doi.org/10.1098/rspb.2019.1992>.
- Du, Z., L. Li, J. Liang, B. Kang, W. Meng, and H. Li. 2024. "National Scale Biodiversity Conservation Priorities Based on Integrated Multiple Vulnerability Features in China." *Ecological Indicators* 160: 111914. <https://doi.org/10.1016/j.ecolind.2024.111914>.
- Ellis, N., S. J. Smith, and C. Roland Pitcher. 2012. "Gradient Forests: Calculating Importance Gradients on Physical Predictors." *Ecology* 93, no. 1: 156–168. <https://doi.org/10.1890/11-0252.1>.
- Elmer, K. R. 2016. "Genomic Tools for New Insights to Variation, Adaptation, and Evolution in the Salmonid Fishes: A Perspective for Charr." *Hydrobiologia* 783, no. 1: 191–208. <https://doi.org/10.1007/s10750-015-2614-5>.
- Esin, E. V., G. N. Markevich, N. O. Melnik, D. V. Zlenko, and F. N. Shkil. 2021. "Ambient Temperature as a Factor Contributing to the Developmental Divergence in Sympatric Salmonids." *PLoS One* 16, no. 10: e0258536. <https://doi.org/10.1371/journal.pone.0258536>.
- Etheridge, E. C., C. E. Adams, C. W. Bean, et al. 2012. "Are Phenotypic Traits Useful for Differentiating Among a Priori *Coregonus* Taxa?" *Journal of Fish Biology* 80, no. 2: 387–407. <https://doi.org/10.1111/j.1095-8649.2011.03189.x>.
- Fang, B., P. Momigliano, K. K. Kahilainen, and J. Merilä. 2022. "Allopatric Origin of Sympatric Whitefish Morphs With Insights on the Genetic Basis of Their Reproductive Isolation." *Evolution* 76, no. 8: 1905–1913. <https://doi.org/10.1111/evo.14559>.
- Fenton, S. 2024. The Adaptive Diversity of Arctic Charr (*Salvelinus alpinus*) in Scotland [Doctoral thesis]. University of Glasgow. <https://doi.org/10.5525/gla.thesis.84449>.
- Fenton, S., C. W. Bean, S. A. M. Martin, S. J. Poultney, C. E. Adams, and K. R. Elmer. 2024. "Ecological Opportunity Leads to Higher Diversity and Probability of Trophic Specialization in Arctic Charr." *Evolutionary Journal of the Linnean Society* 3, no. 1: kzae014. <https://doi.org/10.1093/evolinnean/kzae014>.
- Fenton, S., K. R. Elmer, C. W. Bean, and C. E. Adams. 2023. "How Glaciation Impacted Evolutionary History and Contemporary Genetic Diversity of Flora and Fauna in the British Isles." *Scottish Geographical Journal* 139, no. 3–4: 445–465. <https://doi.org/10.1080/14702541.2023.2231407>.
- Fenton, S., A. Jacobs, C. W. Bean, C. E. Adams, and K. R. Elmer. 2024. "Genomic Underpinnings of Head and Body Shape in Arctic Charr Ecomorph Pairs." *Molecular Ecology* 33, no. 7: e17305. <https://doi.org/10.1111/mec.17305>.
- Fick, S. E., and R. J. Hijmans. 2017. "WorldClim 2: New 1-Km Spatial Resolution Climate Surfaces for Global Land Areas." *International Journal of Climatology* 37, no. 12: 4302–4315. <https://doi.org/10.1002/joc.5086>.
- Finn, D. S., M. T. Bogan, and D. A. Lytle. 2009. "Demographic Stability Metrics for Conservation Prioritization of Isolated Populations." *Conservation Biology* 23, no. 5: 1185–1194. <https://doi.org/10.1111/j.1523-1739.2009.01226.x>.
- Finstad, A. G., and C. L. Hein. 2012. "Migrate or Stay: Terrestrial Primary Productivity and Climate Drive Anadromy in Arctic Char." *Global Change Biology* 18, no. 8: 2487–2497. <https://doi.org/10.1111/j.1365-2486.2012.02717.x>.
- Fischer, J., and D. B. Lindenmayer. 2007. "Landscape Modification and Habitat Fragmentation: A Synthesis." *Global Ecology and Biogeography* 16: 265–280. <https://doi.org/10.1111/j.1466-8238.2006.00287.x>.
- Fitzpatrick, M. C., V. E. Chhatre, R. Y. Soolanayakanahally, and S. R. Keller. 2021. "Experimental Support for Genomic Prediction of Climate Maladaptation Using the Machine Learning Approach Gradient Forests." *Molecular Ecology Resources* 21, no. 8: 2749–2765. <https://doi.org/10.1111/1755-0998.13374>.
- Fitzpatrick, M. C., and S. R. Keller. 2015. "Ecological Genomics Meets Community-Level Modelling of Biodiversity: Mapping the Genomic Landscape of Current and Future Environmental Adaptation." *Ecology Letters* 18, no. 1: 1–16. <https://doi.org/10.1111/ele.12376>.
- Forester, B. R., M. Murphy, C. Mellison, et al. 2022. "Genomics-Informed Delineation of Conservation Units in a Desert Amphibian." *Molecular Ecology* 31, no. 20: 5249–5269. <https://doi.org/10.1111/mec.16660>.
- Frankham, R., J. D. Ballou, and D. A. Briscoe. 2009. *Introduction to Conservation Genetics*. 2nd ed. Cambridge University Press.
- Fraser, D., C. E. Adams, and F. A. Huntingford. 1998. "Trophic Polymorphism Among Arctic Charr *Salvelinus alpinus* L., From Loch Erich, Scotland." *Ecology of Freshwater Fish* 7, no. 4: 184–191. <https://doi.org/10.1111/j.1600-0633.1998.tb00185.x>.
- Funk, W. C., R. E. Lovich, P. A. Hohenlohe, et al. 2016. "Adaptive Divergence Despite Strong Genetic Drift: Genomic Analysis of the Evolutionary Mechanisms Causing Genetic Differentiation in the Island Fox (*Urocyon littoralis*)." *Molecular Ecology* 25, no. 10: 2176–2194. <https://doi.org/10.1111/mec.13605>.
- Funk, W. C., J. K. McKay, P. A. Hohenlohe, and F. W. Allendorf. 2012. "Harnessing Genomics for Delineating Conservation Units." *Trends in Ecology & Evolution* 27, no. 9: 489–496. <https://doi.org/10.1016/j.tree.2012.05.012>.
- Galland, L. M., J. B. Simmons, J. P. Jahner, et al. 2021. "Hierarchical Genetic Structure and Implications for Conservation of the World's Largest Salmonid, *Hucho taimen*." *Scientific Reports* 11, no. 1: 20508. <https://doi.org/10.1038/s41598-021-99530-3>.
- Garduño-Paz, M. V., and C. E. Adams. 2010. "Discrete Prey Availability Promotes Foraging Segregation and Early Divergence in Arctic Charr, *Salvelinus alpinus*." *Hydrobiologia* 650, no. 1: 15–26. <https://doi.org/10.1007/s10750-009-0055-8>.
- Garduño-Paz, M. V., C. E. Adams, E. Verspoor, D. Knox, and C. Harrod. 2012. "Convergent Evolutionary Processes Driven by Foraging Opportunity in Two Sympatric Morph Pairs of Arctic Charr With Contrasting Post-Glacial Origins." *Biological Journal of the Linnean Society* 106, no. 4: 794–806. <https://doi.org/10.1111/j.1095-8312.2012.01906.x>.
- Garduño-Paz, M. V., M. Demetriou, and C. E. Adams. 2010. "Variation in Scale Shape Among Alternative Sympatric Phenotypes of Arctic Charr *Salvelinus alpinus* From Two Lakes in Scotland." *Journal of Fish Biology* 76, no. 6: 1491–1497. <https://doi.org/10.1111/j.1095-8649.2010.02584.x>.
- Gómez, A., and D. H. Lunt. 2007. "Refugia Within Refugia: Patterns of Phylogeographic Concordance in the Iberian Peninsula." In *Phylogeography of Southern European Refugia*, 155–188. Springer Netherlands. https://doi.org/10.1007/1-4020-4904-8_5.
- Gordeeva, N. V., S. S. Alekseyev, A. N. Matveev, and V. P. Samusenok. 2015. "Parallel Evolutionary Divergence in Arctic Char *Salvelinus alpinus* Complex From Transbaikalia: Variation in Differentiation Degree and Segregation of Genetic Diversity Among Sympatric Forms." *Canadian Journal of Fisheries and Aquatic Sciences* 72, no. 1: 96–115. <https://doi.org/10.1139/cjfas-2014-0014>.

- Goudet, J. 2005. "Hierfstat, a Package for R to Compute and Test Hierarchical F-Statistics." *Molecular Ecology Notes* 5: 184–186. <https://doi.org/10.1111/j.1471-8278>.
- Gouthier, L., E. Duval, S. Blanchet, et al. 2023. "Spatial Patterns of Neutral and Functional Genetic Variations Along Dendritic Networks of Riverscape in Brown Trout Populations." *Diversity* 15, no. 6: 784. <https://doi.org/10.3390/d15060784>.
- Graham, C. F., D. R. Boreham, R. G. Manzon, J. Y. Wilson, and C. M. Somers. 2022. "Population Structure of Lake Whitefish (*Coregonus clupeaformis*) From the Mississippian Lineage in North America." *Facets* 7: 853–874. <https://doi.org/10.1139/facets-2021-0191>.
- Gratton, P., G. Allegrucci, V. Sbordoni, and A. Gandolfi. 2014. "The Evolutionary Jigsaw Puzzle of the Surviving Trout (*Salmo trutta* L. complex) Diversity in the Italian Region. A Multilocus Bayesian Approach." *Molecular Phylogenetics and Evolution* 79: 292–304. <https://doi.org/10.1016/j.ympev.2014.06.022>.
- Hay, A. C., J. Sandoval-Castillo, G. M. Cooke, N. L. Chao, and L. B. Beheregaray. 2022. "Riverscape Genomics Clarifies Neutral and Adaptive Evolution in an Amazonian Characin Fish (*Tripottheus albus*)." *Frontiers in Ecology and Evolution* 10: 825406. <https://doi.org/10.3389/fevo.2022.825406>.
- Hess, J. E., J. S. Zandt, A. R. Matala, and S. R. Narum. 2016. "Genetic Basis of Adult Migration Timing in Anadromous Steelhead Discovered Through Multivariate Association Testing." *Proceedings of the Royal Society B: Biological Sciences* 283, no. 1830: 20153064. <https://doi.org/10.1098/rspb.2015.3064>.
- Hijmans, R. J., and J. van Etten. 2012. "raster: Geographic Analysis and Modeling With Raster Data. (3.5–15)." <http://CRAN.R-project.org/package=raster>.
- Hoban, S., C. Hvilson, A. Aissi, et al. 2025. "How Can Biodiversity Strategy and Action Plans Incorporate Genetic Diversity and Align With Global Commitments?" *Bioscience* 75, no. 1: 47–60. <https://doi.org/10.1093/biosci/biae106>.
- Hoelzel, A. R. 2023. "Where to Now With the Evolutionarily Significant Unit?" *Trends in Ecology & Evolution* 38, no. 12: 1134–1142. <https://doi.org/10.1016/j.tree.2023.07.005>.
- Hollingsworth, P. M., D. O'Brien, R. A. Ennos, et al. 2019. "Conserving Genetic Diversity: Development of a National Approach for Addressing Aichi Biodiversity Target 13 That Includes Wild Species." <https://www.nature.scot/doc/scotlands-biodiversity-progress-2020-aichi-targets-aichi-target-13-genetic-diversity-maintained>.
- Hooker, O. E., C. E. Adams, and L. Chavarie. 2023. "Arctic Charr Phenotypic Responses to Abrupt Generational Scale Temperature Change: An Insight Into How Cold-Water Fish Could Respond to Extreme Climatic Events." *Environmental Biology of Fishes* 106, no. 5: 909–922. <https://doi.org/10.1007/s10641-022-01363-0>.
- Hooker, O. E., J. Barry, T. E. Van Leeuwen, et al. 2016. "Morphological, Ecological and Behavioural Differentiation of Sympatric Profundal and Pelagic Arctic Charr (*Salvelinus alpinus*) in Loch Dughail Scotland." *Hydrobiologia* 783, no. 1: 209–221. <https://doi.org/10.1007/s10750-015-2599-0>.
- Isaak, D. J., and M. K. Young. 2023. "Cold-Water Habitats, Climate Refugia, and Their Utility for Conserving Salmonid Fishes." *Canadian Journal of Fisheries and Aquatic Sciences* 80, no. 7: 1187–1206. <https://doi.org/10.1139/cjfas-2022-0302>.
- Jacobs, A. 2018. "The Population Genomic Origins of Ecological Specialisation in Salmonid Fishes." [Doctoral Dissertation]. University of Glasgow. <https://doi.org/10.5525/gla.thesis.30678>.
- Jacobs, A., M. Carruthers, A. Yurchenko, et al. 2020. "Parallelism in Eco-Morphology and Gene Expression Despite Variable Evolutionary and Genomic Backgrounds in a Holarctic Fish." *PLoS Genetics* 16, no. 4: e1008658. <https://doi.org/10.1371/journal.pgen.1008658>.
- Jacobs, A., R. Womack, M. Chen, K. Gharbi, and K. R. Elmer. 2017. "Significant Synteny and Colocalization of Ecologically Relevant Quantitative Trait Loci Within and Across Species of Salmonid Fishes." *Genetics* 207, no. 2: 741–754. <https://doi.org/10.1534/genetics.117.300093>.
- Jacobsen, M. W., N. W. Jensen, R. Nygaard, et al. 2022. "A Melting Pot in the Arctic: Analysis of Mitogenome Variation in Arctic Char (*Salvelinus alpinus*) Reveals a 1000-Km Contact Zone Between Highly Divergent Lineages." *Ecology of Freshwater Fish* 31, no. 2: 330–346. <https://doi.org/10.1111/eff.12633>.
- Jombart, T. 2008. "Adegenet: A R Package for the Multivariate Analysis of Genetic Markers." *Bioinformatics* 24, no. 11: 1403–1405. <https://doi.org/10.1093/bioinformatics/btn129>.
- Jonsson, B., and N. Jonsson. 2001. "Polymorphism and Speciation in Arctic Charr." *Journal of Fish Biology* 58, no. 3: 605–638. <https://doi.org/10.1006/jfbi.2000.1515>.
- Jørgensen, E. H., and H. K. Johnsen. 2014. "Rhythmic Life of the Arctic Charr: Adaptations to Life at the Edge." *Marine Genomics* 14: 71–81. <https://doi.org/10.1016/j.margen.2013.10.005>.
- Kamvar, Z. N., J. C. Brooks, and N. J. Grünwald. 2015. "Novel R Tools for Analysis of Genome-Wide Population Genetic Data With Emphasis on Clonality." *Frontiers in Genetics* 6: 208. <https://doi.org/10.3389/fgene.2015.00208>.
- Kamvar, Z. N., M. M. López-Urbe, S. Coughlan, N. J. Grünwald, H. Lapp, and S. Manel. 2017. "Developing Educational Resources for Population Genetics in R: An Open and Collaborative Approach." *Molecular Ecology Resources* 17, no. 1: 120–128. <https://doi.org/10.1111/1755-0998.12558>.
- Kazyak, D. C., B. A. Lubinski, M. A. Kulp, et al. 2022. "Population Genetics of Brook Trout in the Southern Appalachian Mountains." *Transactions of the American Fisheries Society* 151, no. 2: 127–149. <https://doi.org/10.1002/tafs.10337>.
- Kelly, S., T. N. Moore, E. de Eyto, et al. 2020. "Warming Winters Threaten Peripheral Arctic Charr Populations of Europe." *Climatic Change* 163, no. 1: 599–618. <https://doi.org/10.1007/s10584-020-02887-z>.
- Kess, T., J. B. Dempson, S. J. Lehnert, et al. 2021. "Genomic Basis of Deep-Water Adaptation in Arctic Charr (*Salvelinus alpinus*) Morphs." *Molecular Ecology* 30, no. 18: 4415–4432. <https://doi.org/10.1111/mec.16033>.
- Kettle-White, A. 2001. "New Fish Records: Stoneloach, Roach and a Spring Spawning Arctic Charr Population, Loch Awe, Argyll." *Glasgow Naturalist* 23, no. 6: 120.
- Kitada, S., R. Nakamichi, and H. Kishino. 2021. "Understanding Population Structure in an Evolutionary Context: Population-Specific *F_{ST}* and Pairwise *F_{ST}*." *G3 (Bethesda, Md.)* 11, no. 11: jkab316. <https://doi.org/10.1093/g3journal/jkab316>.
- Klemetsen, A. 2010. "The Charr Problem Revisited: Exceptional Phenotypic Plasticity Promotes Ecological Speciation in Postglacial Lakes." *Freshwater Reviews* 3, no. 1: 49–74. <https://doi.org/10.4290/frj-3.1.3>.
- Knudsen, R., A. Klemetsen, S. Alekseyev, C. E. Adams, and M. Power. 2016. "The Role of *Salvelinus* in Contemporary Studies of Evolution, Trophic Ecology and Anthropogenic Change." *Hydrobiologia* 783, no. 1: 1–9. <https://doi.org/10.1007/s10750-016-2999-9>.
- Kottelat, M., and J. Freyhof. 2007. *Handbook of European Freshwater Fishes*. Self-Published.
- Kristjánsson, B. K., C. A. L. Leblanc, S. Skúlason, S. S. Snorrason, and D. L. G. Noakes. 2018. "Phenotypic Plasticity in the Morphology of Small Benthic Icelandic Arctic Charr (*Salvelinus alpinus*)." *Ecology of Freshwater Fish* 27, no. 3: 636–645. <https://doi.org/10.1111/eff.12380>.
- Kristjánsson, B. K., S. Skúlason, S. S. Snorrason, and D. L. G. Noakes. 2012. "Fine-Scale Parallel Patterns in Diversity of Small Benthic

- Arctic Charr (*Salvelinus alpinus*) in Relation to the Ecology of Lava/ Groundwater Habitats." *Ecology and Evolution* 2, no. 6: 1099–1112. <https://doi.org/10.1002/ece3.235>.
- Láruson, Á. J., M. C. Fitzpatrick, S. R. Keller, B. C. Haller, and K. E. Lotterhos. 2022. "Seeing the Forest for the Trees: Assessing Genetic Offset Predictions From Gradient Forest." *Evolutionary Applications* 15, no. 3: 403–416. <https://doi.org/10.1111/eva.13354>.
- Lavictoire, L., C. W. Bean, B. Hänfling, D. Ottewell, S. Thackeray, and M. Palin. 2025. *Species on the Edge: Arctic Charr at the Southern Extent of Its Range*. Freshwater Biological Association.
- Layton, K. K. S., P. V. R. Snelgrove, J. B. Dempson, et al. 2021. "Genomic Evidence of Past and Future Climate-Linked Loss in a Migratory Arctic Fish." *Nature Climate Change* 11, no. 2: 158–165. <https://doi.org/10.1038/s41558-020-00959-7>.
- Le Coeur, C., J. Storkey, and S. Ramula. 2021. "Population Responses to Observed Climate Variability Across Multiple Organismal Groups." *Oikos* 130, no. 3: 476–487. <https://doi.org/10.1111/oik.07371>.
- Lema, S. C., S. L. Bock, M. M. Malley, and E. A. Elkins. 2019. "Warming Waters Beget Smaller Fish: Evidence for Reduced Size and Altered Morphology in a Desert Fish Following Anthropogenic Temperature Change." *Biology Letters* 15, no. 10: 20190518. <https://doi.org/10.1098/rsbl.2019.0518>.
- Lesica, P., and F. W. Allendorf. 1995. "When Are Peripheral Populations Valuable for Conservation?" *Conservation Biology* 9, no. 4: 753–760. <https://doi.org/10.1046/j.1523-1739.1995.09040753.x>.
- Li, H. 2013. "Aligning Sequence Reads, Clone Sequences and Assembly Contigs With BWA-MEM."
- Lyle, A. A., P. S. Maitland, and I. J. Winfield. 2006. "Translocation of Vendace From Derwent Water to Safe Refuge Locations 2005/2006 Final Report."
- Mable, B. K. 2019. "Conservation of Adaptive Potential and Functional Diversity: Integrating Old and New Approaches." *Conservation Genetics* 20, no. 1: 89–100. <https://doi.org/10.1007/s10592-018-1129-9>.
- Mainstone, C. P., R. A. Hall, T. W. Hatton-Ellis, P. J. Boon, C. W. Bean, and A. S. L. Lee. 2018. "Guidelines for the Selection of Biological SSSIs." Part 2: Detailed Guidelines for Habitats and Species Groups. Chapter 6 Freshwater Habitats. Joint Nature Conservation Committee.
- Maitland, P. S., and C. E. Adams. 2018. *Arctic Charr in the Lochs of Scotland: An Assessment of Distribution and Status*. Fast-Print Publishing.
- Maitland, P. S., L. May, D. H. Jones, and C. R. Doughty. 1991. "Ecology and Conservation of Arctic Charr, *Salvelinus alpinus* (L.), in Loch Doon, an Acidifying Loch in Southwest Scotland." *Biological Conservation* 55, no. 2: 167–197. [https://doi.org/10.1016/0006-3207\(91\)90055-E](https://doi.org/10.1016/0006-3207(91)90055-E).
- Maitland, P. S., I. J. Winfield, I. D. McCarthy, and F. Igoe. 2007. "The Status of Arctic Charr *Salvelinus alpinus* in Britain and Ireland." *Ecology of Freshwater Fish* 16, no. 1: 6–19. <https://doi.org/10.1111/j.1600-0633.2006.00167.x>.
- Malmquist, H. J., S. S. Snorrason, S. Skúlason, B. Jonsson, O. T. Sandlund, and P. M. Jonasson. 1992. "Diet Differentiation in Polymorphic Arctic Charr in Thingvallavatn, Iceland." *Journal of Animal Ecology* 61, no. 1: 21–35.
- Mastretta-Yanes, A., N. Arrigo, N. Alvarez, T. H. Jorgensen, D. Piñero, and B. C. Emerson. 2015. "Restriction Site-Associated DNA Sequencing, Genotyping Error Estimation and de Novo Assembly Optimization for Population Genetic Inference." *Molecular Ecology Resources* 15, no. 1: 28–41. <https://doi.org/10.1111/1755-0998.12291>.
- Matala, A. P., M. W. Ackerman, M. R. Campbell, and S. R. Narum. 2014. "Relative Contributions of Neutral and Non-Neutral Genetic Differentiation to Inform Conservation of Steelhead Trout Across Highly Variable Landscapes." *Evolutionary Applications* 7, no. 6: 682–701. <https://doi.org/10.1111/eva.12174>.
- McLeod, C., M. Yeo, A. Brown, A. Burn, J. Hopkins, and S. Way. 2009. "The Habitats Directive: Selection of Special Areas of Conservation in the UK, 2nd Edition (Version 4)." Joint Nature Conservation Committee.
- Mee, J. A., L. Bernatchez, J. D. Reist, S. M. Rogers, and E. B. Taylor. 2015. "Identifying Designatable Units for Intraspecific Conservation Prioritization: A Hierarchical Approach Applied to the Lake Whitefish Species Complex (*Coregonus* spp.)." *Evolutionary Applications* 8, no. 5: 423–441. <https://doi.org/10.1111/eva.12247>.
- Miller, C. V., C. M. Bossu, J. F. Saracco, et al. 2024. "Genomics-Informed Conservation Units Reveal Spatial Variation in Climate Vulnerability in a Migratory Bird." *Molecular Ecology* 33, no. 1: e17199. <https://doi.org/10.1111/mec.17199>.
- Minter, M., D. O'Brien, J. Cottrell, R. Ennos, J. K. Hill, and J. Hall. 2021. "Exploring the Potential for 'Gene Conservation Units' to Conserve Genetic Diversity in Wild Populations." *Ecological Solutions and Evidence* 2, no. 2: e12061. <https://doi.org/10.1002/2688-8319.12061>.
- Mocetti, P., J. R. Dodd, D. A. Joyce, A. D. Nunn, B. Gillespie, and J. D. Bolland. 2024. "Genetic Consequences of Improved River Connectivity in Brown Trout (*Salmo trutta* L.)." *Evolutionary Applications* 17, no. 4: e13660. <https://doi.org/10.1111/eva.13660>.
- Mocetti, P., A. Siwertsson, R. Kjær, et al. 2019. "Contrasting Patterns in Trophic Niche Evolution of Polymorphic Arctic Charr Populations in Two Subarctic Norwegian Lakes." *Hydrobiologia* 840, no. 1: 281–299. <https://doi.org/10.1007/s10750-019-3969-9>.
- Moore, J. S., R. Bajno, J. D. Reist, and E. B. Taylor. 2015. "Post-Glacial Recolonization of the North American Arctic by Arctic Char (*Salvelinus alpinus*): Genetic Evidence of Multiple Northern Refugia and Hybridization Between Glacial Lineages." *Journal of Biogeography* 42, no. 11: 2089–2100. <https://doi.org/10.1111/jbi.12600>.
- Moritz, C. 1994. "Defining 'Evolutionarily Significant Units' for Conservation." *Trends in Ecology & Evolution* 9, no. 10: 373–375. [https://doi.org/10.1016/0169-5347\(94\)90057-4](https://doi.org/10.1016/0169-5347(94)90057-4).
- Muir, C. A., S. R. Garner, S. Damjanovski, and B. D. Neff. 2022. "Temperature-Dependent Plasticity Mediates Heart Morphology and Thermal Performance of Cardiac Function in Juvenile Atlantic Salmon (*Salmo salar*)." *Journal of Experimental Biology* 225, no. 16: jeb244305. <https://doi.org/10.1242/jeb.244305>.
- Murray, J., and L. Pullar. 1910. *Bathymetrical Survey of the Scottish Fresh Water Lochs*. Vol. 6. Challenger Office.
- Mussmann, S. M., M. R. Douglas, D. D. Oakey, and M. E. Douglas. 2020. "Defining Relictual Biodiversity: Conservation Units in Speckled Dace (*Leuciscidae: Rhinichthys osculus*) of the Greater Death Valley Ecosystem." *Ecology and Evolution* 10, no. 19: 10798–10817. <https://doi.org/10.1002/ece3.6736>.
- National River Flow Archive. 2014. "Integrated Hydrological Units of the United Kingdom: Hydrometric Areas With Coastline." NERC Environmental Information Data Centre. <https://doi.org/10.5285/1957166d-7523-44f4-b279-aa5314163237>.
- Öhlund, G., M. Bodin, K. A. Nilsson, et al. 2020. "Ecological Speciation in European Whitefish Is Driven by a Large-Gaped Predator." *Evolution Letters* 4, no. 3: 243–256. <https://doi.org/10.1002/evl3.167>.
- Oksanen, J., G. Simpson, F. Blanchet, et al. 2022. "vegan: Community Ecology Package (2.5–7)." <https://cran.r-project.org/package=vegan>.
- Palsbøll, P., M. Bérubé, and F. Allendorf. 2007. "Identification of Management Units Using Population Genetic Data." *Trends in Ecology & Evolution* 22, no. 1: 11–16. <https://doi.org/10.1016/j.tree.2006.09.003>.
- Prince, D. J., S. M. O'Rourke, T. Q. Thompson, et al. 2017. "The Evolutionary Basis of Premature Migration in Pacific Salmon Highlights the Utility of Genomics for Informing Conservation." *Science Advances* 3, no. 8: e1603198. <https://doi.org/10.1126/sciadv.1603198>.

- Quinlan, A. R., and I. M. Hall. 2010. "BEDTools: A Flexible Suite of Utilities for Comparing Genomic Features." *Bioinformatics* 26, no. 6: 841–842. <https://doi.org/10.1093/bioinformatics/btq033>.
- R Core Team. 2021. "R: A Language and Environment for Statistical Computing." R Foundation for Statistical Computing. <https://www.R-project.org/>.
- Ratner, B. 2009. "The Correlation Coefficient: Its Values Range Between 1/1, or Do They?" *Journal of Targeting, Measurement and Analysis for Marketing* 17, no. 2: 139–142. <https://doi.org/10.1057/jt.2009.5>.
- Recknagel, H., O. E. Hooker, C. E. Adams, and K. R. Elmer. 2017. "Ecosystem Size Predicts Eco-Morphological Variability in a Postglacial Diversification." *Ecology and Evolution* 7, no. 15: 5560–5570. <https://doi.org/10.1002/ece3.3013>.
- Recknagel, H., A. Jacobs, P. Herzyk, and K. R. Elmer. 2015. "Double-Digest RAD Sequencing Using Ion Proton Semiconductor Platform (ddRADseq) With Nonmodel Organisms." *Molecular Ecology Resources* 15, no. 6: 1316–1329. <https://doi.org/10.1111/1755-0998.12406>.
- Reed, D. H., and R. Frankham. 2003. "Correlation Between Fitness and Genetic Diversity." *Conservation Biology* 17, no. 1: 230–237.
- Rellstab, C., B. Dauphin, and M. Exposito-Alonso. 2021. "Prospects and Limitations of Genomic Offset in Conservation Management." *Evolutionary Applications* 14, no. 5: 1202–1212. <https://doi.org/10.1111/eva.13205>.
- Reusch, T. B. H., A. Ehlers, A. Hä, and B. Worm. 2005. "Ecosystem Recovery After Climatic Extremes Enhanced by Genotypic Diversity." *PNAS* 102: 2826–2831.
- Riesch, R., R. A. Martin, S. E. Diamond, J. Jourdan, M. Plath, and R. Brian Langerhans. 2018. "Thermal Regime Drives a Latitudinal Gradient in Morphology and Life History in a Livebearing Fish." *Biological Journal of the Linnean Society* 125, no. 1: 126–141. <https://doi.org/10.1093/biolinnean/bly095>.
- Robertson, J. M., K. M. Langin, T. S. Sillett, S. A. Morrison, C. K. Ghalambor, and W. C. Funk. 2014. "Identifying Evolutionarily Significant Units and Prioritizing Populations for Management on Islands." *Monographs of the Western North American Naturalist* 7, no. 1: 397–411. <https://doi.org/10.3398/042.007.0130>.
- Rochette, N. C., A. G. Rivera-Colón, and J. M. Catchen. 2019. "Stacks 2: Analytical Methods for Paired-End Sequencing Improve RADseq-Based Population Genomics." *Molecular Ecology* 28, no. 21: 4737–4754. <https://doi.org/10.1111/mec.15253>.
- Rodger, J. R., H. M. Honkanen, C. R. Bradley, P. Boylan, P. A. Prodöhl, and C. E. Adams. 2021. "Genetic Structuring Across Alternative Life-History Tactics and Small Spatial Scales in Brown Trout (*Salmo trutta*)." *Ecology of Freshwater Fish* 30, no. 2: 174–183. <https://doi.org/10.1111/eff.12573>.
- Ryder, O. A. 1986. "Species Conservation and Systematics: The Dilemma of Subspecies." *Trends in Ecology & Evolution* 1, no. 1: 9–10. [https://doi.org/10.1016/0169-5347\(86\)90059-5](https://doi.org/10.1016/0169-5347(86)90059-5).
- Salisbury, S. J., R. Perry, D. Keefe, et al. 2023. "Geography, Environment, and Colonization History Interact With Morph Type to Shape Genomic Variation in an Arctic Fish." *Molecular Ecology* 32, no. 12: 3025–3043. <https://doi.org/10.1111/mec.16913>.
- Salisbury, S. J., and D. E. Ruzzante. 2022. "Genetic Causes and Consequences of Sympatric Morph Divergence in Salmonidae: A Search for Mechanisms." *Annual Review of Animal Biosciences* 10, no. 1: 81–106. <https://doi.org/10.1146/annurev-animal-051021-080709>.
- Schindler, D. E., R. Hilborn, B. Chasco, et al. 2010. "Population Diversity and the Portfolio Effect in an Exploited Species." *Nature* 465, no. 7298: 609–612. <https://doi.org/10.1038/nature09060>.
- Schluter, D. 1996. "Ecological Speciation in Postglacial Fishes." *Philosophical Transactions of the Royal Society, B: Biological Sciences* 351, no. 1341: 807–814. <https://doi.org/10.1098/rstb.1996.0075>.
- Schneider, C. A., W. S. Rasband, and K. W. Eliceiri. 2012. "NIH Image to ImageJ: 25 Years of Image Analysis." *Nature Methods* 9, no. 7: 671–675. <https://doi.org/10.1038/nmeth.2089>.
- SEPA. 2021. "The River Basin Management Plan for Scotland 2021–2027."
- Sgrò, C. M., A. J. Lowe, and A. A. Hoffmann. 2011. "Building Evolutionary Resilience for Conserving Biodiversity Under Climate Change." *Evolutionary Applications* 4, no. 2: 326–337. <https://doi.org/10.1111/j.1752-4571.2010.00157.x>.
- Shaney, K. J., L. G. Diaz-Ramirez, S. Espindola, S. Castañeda-Rico, V. Berovides-Álvarez, and E. Vázquez-Domínguez. 2020. "Defining Intraspecific Conservation Units in the Endemic Cuban Rock Iguanas (*Cyclura nubila nubila*)." *Scientific Reports* 10, no. 1: 21607. <https://doi.org/10.1038/s41598-020-78664-w>.
- Shikano, T., A. Järvinen, P. Marjamäki, K. K. Kahilainen, and J. Merilä. 2015. "Genetic Variability and Structuring of Arctic Charr (*Salvelinus alpinus*) Populations in Northern Fennoscandia." *PLoS One* 10, no. 10: e0140344. <https://doi.org/10.1371/journal.pone.0140344>.
- Skúlason, S., S. S. Snorrason, D. L. G. Noakes, and M. M. Ferguson. 1996. "Genetic Basis of Life History Variations Among Sympatric Morphs of Arctic Char *Salvelinus alpinus*." *Canadian Journal of Fisheries and Aquatic Sciences* 53, no. 8: 1807–1813. <https://doi.org/10.1139/cjfas-53-8-1807>.
- Supek, F., M. Bošnjak, N. Škunca, and T. Šmuc. 2011. "Revigo Summarizes and Visualizes Long Lists of Gene Ontology Terms." *PLoS One* 6, no. 7: e21800. <https://doi.org/10.1371/journal.pone.0021800>.
- Tiberti, R., and A. Splendiani. 2019. "Management of a Highly Unlikely Native Fish: The Case of Arctic Charr *Salvelinus alpinus* From the Southern Alps." *Aquatic Conservation: Marine and Freshwater Ecosystems* 29, no. 2: 312–320. <https://doi.org/10.1002/aqc.3027>.
- Tiddy, I. C., K. Schneider, and K. R. Elmer. 2024. "Environmental Correlates of Adaptive Diversification in Postglacial Freshwater Fishes." *Journal of Fish Biology* 104, no. 3: 517–535. <https://doi.org/10.1111/jfb.15621>.
- Tigano, A., T. Weir, H. G. M. Ward, et al. 2024. "Genomic Vulnerability of a Freshwater Salmonid Under Climate Change." *Evolutionary Applications* 17, no. 2: e13602. <https://doi.org/10.1111/eva.13602>.
- Vavalidis, T., S. Zogaris, A. N. Economou, A. S. Kallimanis, and D. C. Bobori. 2019. "Changes in Fish Taxonomy Affect Freshwater Biogeographical Regionalisations: Insights From Greece." *Water* 11, no. 9: 1743. <https://doi.org/10.3390/w11091743>.
- Verspoor, E., D. Knox, R. Greer, and J. Hammar. 2010. "Mitochondrial DNA Variation in Arctic Charr (*Salvelinus alpinus* (L.)) Morphs From Loch Rannoch, Scotland: Evidence for Allopatric and Peripatric Divergence." *Hydrobiologia* 650, no. 1: 117–131. <https://doi.org/10.1007/s10750-010-0106-1>.
- Walker, A. F. 2007. "Stream Spawning of Arctic Charr in Scotland." *Ecology of Freshwater Fish* 16, no. 1: 47–53. <https://doi.org/10.1111/j.1600-0633.2006.00164.x>.
- Wang, J. 2022. "Fast and Accurate Population Admixture Inference From Genotype Data From a Few Microsatellites to Millions of SNPs." *Heredity* 129, no. 2: 79–92. <https://doi.org/10.1038/s41437-022-00535-z>.
- Waples, R. S. 1991. "Pacific Salmon, *Oncorhynchus* spp., and the Definition of 'Species' Under the Endangered Species Act." *Marine Fisheries Review* 53, no. 3: 11–22.
- Waples, R. S., M. J. Ford, K. Nichols, et al. 2022. "Implications of Large-Effect Loci for Conservation: A Review and Case Study With Pacific Salmon." *Journal of Heredity* 113, no. 2: 121–144. <https://doi.org/10.1093/jhered/esab069>.
- Waples, R. S., and S. T. Lindley. 2018. "Genomics and Conservation Units: The Genetic Basis of Adult Migration Timing in Pacific

Salmonids." *Evolutionary Applications* 11, no. 9: 1518–1526. <https://doi.org/10.1111/eva.12687>.

Wilson, A. J., D. Gíslason, S. Skúlason, et al. 2004. "Population Genetic Structure of Arctic Charr, *Salvelinus alpinus* From Northwest Europe on Large and Small Spatial Scales." *Molecular Ecology* 13, no. 5: 1129–1142. <https://doi.org/10.1111/j.1365-294X.2004.02149.x>.

Winfield, I. J., J. M. Fletcher, and J. B. James. 2008. "The Arctic Charr (*Salvelinus alpinus*) Populations of Windermere, UK: Population Trends Associated With Eutrophication, Climate Change and Increased Abundance of Roach (*Rutilus rutilus*)." *Environmental Biology of Fishes* 83, no. 1: 25–35. <https://doi.org/10.1007/s10641-007-9235-4>.

Winfield, I. J., J. Hateley, J. M. Fletcher, J. B. James, C. W. Bean, and P. Claburn. 2010. "Population Trends of Arctic Charr (*Salvelinus alpinus*) in the UK: Assessing the Evidence for a Widespread Decline in Response to Climate Change." *Hydrobiologia* 650, no. 1: 55–65. <https://doi.org/10.1007/s10750-009-0078-1>.

Zhao, T., W. Zhang, J. Zhou, et al. 2020. "Niche Divergence of Evolutionarily Significant Units With Implications for Repopulation Programs of the World's Largest Amphibians." *Science of the Total Environment* 738: 140269. <https://doi.org/10.1016/j.scitotenv.2020.140269>.

Supporting Information

Additional supporting information can be found online in the Supporting Information section. **Figure S1:** Plots showing the values of two estimators DLK2 and FSTIS estimates across the various K (number of genetic clusters) tested in admixture analysis to determine the most likely number of K. **Figure S2:** Violin plot of Expected Heterozygosity, Nucleotide Diversity and Observed Heterozygosity ranges for all populations. Black dot represents the mean values. **Figure S3:** Redundancy analysis plot showing the directions each uncorrelated variables pull in across RDA1 and RDA2. Colours represent each of the individual populations in the dataset. **Figure S4:** R-squared importance plots for uncorrelated environmental variables in explaining (A) adaptive variance ($N = 1,104$ SNPs) and (B) variance of climate associated SNPs used for genetic offset analysis ($N = 235$ SNPs). **Figure S5:** Principal component analysis run on the 10 uncorrelated environmental variables used to identify the putatively adaptive SNPs. (A) shows the placements of each lake of origin on PC1 and PC2 while (B) shows the corresponding loadings of each of the environmental variables. **Figure S6:** EV1 vs EV2 from PCA using SNPs ($N = 1087$) identified from a redundancy analysis when using admixture proportions to control for neutral structure. Individuals are coloured by lake of origin. **Figure S7:** EV1 vs EV2 from PCA using SNPs ($N = 1104$) identified from a redundancy analysis run using population level allele frequencies rather than individual SNP calls. Individuals are coloured by lake of origin. **Figure S8:** Plot of geographic distribution of populations. **Figure S9:** Relationship between genetic offset scores and levels of observed heterozygosity. Populations in the same lake of origin (i.e., ecotype pairs) are given the same offset score. **Figure S10:** Genetic offset scores vs latitude under two RCP scenarios, RCP 4.5 (in blue) and RCP 8.5 (orange). Each point represents a different lake of origin. **Figure S11:** The relationship between genetic offset and distance to sea when only using populations with five or more individuals when calculating genetic offset. **Table S1:** A table of bathymetric (maximum depth, mean depth, surface area and littoral zone percentage), altitude and climatic (Bio1-19) data for all lakes. Note populations in the same lake are given the same values and so values are just provided for the lakes. **Table S2:** List of the 19 climatic variables and 5 bathymetric variables used in the study. **Table S3:** A correlation matrix between all our environmental (altitude, bathymetric and climatic) variables. **Table S4:** Scoring system used to calculate lake sensitivity scores based on lake surface area, depth, and altitude. **Table S5:** A table of pairwise F_{ST} comparisons between all populations in our dataset. **Table S6:** Hierarchical AMOVA results showing the variance explained by various groupings including the east-west drainage divide seen in the neighbour-joining tree and hydrometric area of origin. **Table S7:** ANOVA results showing the significance of each constrained axis in

the RDA in explaining variance. **Table S8:** A table providing the chromosome and position of all 1104 putatively adaptive SNPs. **Table S9:** Number of putatively adaptive SNPs ($N = 1,104$) each uncorrelated environmental variables was the strongest predictor for in the Redundancy Analysis. **Table S10:** A table providing a list of genes which contained putatively adaptive SNPs. Chromosome and position of the SNPs are provided as well the start and end position of the gene and the gene name in the genome assembly. As we included SNPs within 1kbp of the gene, the relative position of the SNP to the gene is provided with a score of zero meaning the SNP is within the gene region. **Table S11:** A table of significant GO terms from our overrepresentation analysis. GO term names and IDs are provided as well as the number of IDs for this term, the number that would be expected, the number we had in our dataset and the associated weighted Fisher's p value. **Table S12:** List of QTLs from Fenton, Jacobs, et al. 2024 that contained adaptive SNPs we identified. **Table S13:** Correlation matrix between the different metrics of population vulnerability: genetic offset scores, lake sensitivity scores, and observed heterozygosity.