Population structure of eulachon *Thaleichthys pacificus* from Northern California to Alaska using single nucleotide polymorphisms from direct amplicon sequencing

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ABSTRACT

Eulachon *Thaleichthys pacificus* is a culturally and ecologically important anadromous smelt (Family Osmeridae) that ranges from Northern California to the southeast Bering Sea. In recent decades, some populations have experienced declines. There is a need to better understand the population structure of this species and develop a method for high-throughput genotyping with the potential for application in genetic stock identification. Here we use a contig-level genome assembly combined with previously published RADseq-derived markers to construct an amplicon panel and genotype 2,125 samples at 521 variant loci (i.e., single marker per amplicon) from 19 populations. Populations with fewer than 35 samples did not cluster as expected in dendrograms (n = 5 populations), suggesting a lower limit on sample size per population with this panel and species. The retained 14 populations ranged from Northern California through Central Alaska, and included a total of 1,989 individuals with an average of 142 individuals per population. Consistent with prior genetic studies, the strongest separation was viewed among three main regions: from Northern California up to and including the Fraser River; north of the Fraser River to southeast Alaska; and within the Gulf of Alaska. Finer-scale resolution of the Fraser River from US populations in mixed-stock analysis may be possible within the southern grouping, as well as additional substructure within the central coast; this will be important to address in future work using simulated and empirical mixed-stock samples. Annual variation was generally low, and populations were often grouped together when separated by year. The observed genetic variation fits a hierarchical island model with strong evidence for isolation-by-distance (IBD) as shown by a comparison of F_{ST} and physical distance between rivers (adj. $R^2 = 0.71$). The results from the SNP panel, even with fewer samples overall, outperformed the previous microsatellite panel (n = 14 markers), showing improved grouping by geography as per *a priori* expectations, higher bootstrap support in dendrograms, and increased evidence of adherence to IBD. This baseline will provide the foundation for future mixed-stock analyses of eulachon, and provides insight into the population structure and further refinement of the appropriate units for management and conservation.

Keywords: Amplicon sequencing; Eulachon; Forage fish; Genotyping; Population Genetics

INTRODUCTION

Eulachon *Thaleichthys pacificus* is a culturally and ecologically important anadromous smelt (Family Osmeridae)
distributed in North America from Northern California to the southeast Bering Sea (Hay and McCarter 2000).
Historically, approximately 95 rivers were considered to have spawning populations along the Northwest Pacific
Coast, with large spawning populations in the Columbia River (USA) and the Fraser River (Canada) (COSEWIC
2011; Moody and Pitcher 2010). Eulachon are an important prey species for birds, marine mammals and fishes, in
part due to a high energetic benefit to cost ratio during foraging (Marston et al. 2002), as well as due to their returning
to spawn when other prey species are scarce (Moody and Pitcher 2010). Eulachon are highly important to First
Nations and American Indian indigenous peoples for both cultural and nutritional purposes, for example through
the preparation and use of the rendered oil from adult eulachon, commonly known as "grease" (Moody and Pitcher
2010). Conservation and management of this species are therefore important goals in both Canada (DFO 2020) and
the US (NMFS 2017).

Declines in southern eulachon populations have been reported coastwide since the mid 1990s, although
some rivers maintain healthy returns including rivers in central Alaska (Ormseth 2018) and northern British
Columbia (BC) (Hay and McCarter 2000). Adjacent regions both in southeast Alaska (SEAK) and Canada have
declined in recent years (COSEWIC 2011). Although biomass has increased overall in Alaska eulachon populations,
some specific runs in the area have had large reductions in returns (Flannery et al. 2013; Ormseth et al. 2008). In a
meta-analysis by Moody and Pitcher (2010), factors were identified as associated with declines including bycatch
in shrimp and hake fisheries, seal and sea lion predation, and increasing sea surface temperature. In Canada, no
single threat could be identified for declines in abundance, although "mortality associated with coastwide changes
in climate, fishing (direct and bycatch) and marine predation were considered to be greater threats… than changes
in habitat or predation within spawning rivers" (DFO 2015; Schweigert et al. 2012). In addition, forage fish
population fluctuations are also dependent on environmental conditions, serving as the main link between climatic
effects on primary producers and higher trophic levels (Guénette et al. 2014; Pikitch et al. 2014). In general, these
fluctuations in forage fish populations are considered to be exacerbated by other factors, such as commercial fishing
(Essington et al. 2015b), although the extent of this is debated (Essington et al. 2015a; Szuwalski and Hilborn 2015).

Within Canada, the Committee on the Status of Endangered Wildlife in Canada (COSEWIC) considers
eulachon to be in three Designatable Units (DUs): the Fraser River; the Central Pacific Coast; and the Nass/Skeena
River (COSEWIC 2011). Both the Fraser River and the Central Pacific Coast DUs were assessed as endangered in
2011 (COSEWIC 2011), whereas the Nass/Skeena DU has been assessed as Special Concern (COSEWIC 2013).
The Fraser River DU and the Central Pacific Coast DU remain under consideration for listing as endangered under
Canada's Species at Risk Act (SARA). Fisheries limitations and bycatch rules in Canada are outlined in Fisheries
Management Plans (e.g., DFO 2020). Within the contiguous US, eulachon are designated as threatened under the
Endangered Species Act within the southern Distinct Population Segment (DPS), ranging from northern California
to the Skeena River, Canada (Gustafson et al. 2012; NMFS 2010). Improved characterization of population structure

and the development of methods to genotype unknown origin eulachon at high-throughput will help to better
understand and manage this species, and will be useful in determining causes of declines.

In general, marine species with large population sizes are expected to have low population differentiation
due to several factors including low effect of drift and high migration rate (Gagnaire et al. 2015) although this is not
always the case, with molecular genetics now providing new resolution to previously considered homogenous stocks
(Hauser and Carvalho 2008). Eulachon are less differentiated among populations than are other anadromous species
such as salmonids (Candy et al. 2015), but are more differentiated than many marine species (Hauser and Carvalho
2008; Waples 1998). This may be due to a shorter in-river residence time, rapid downriver transport after hatch
(Beacham et al. 2005; COSEWIC 2011; McLean and Taylor 2001), and thus less opportunity for imprinting, which
may result in lower natal stream fidelity (Hay and McCarter 2000). Prior studies in eulachon using mitochondrial
DNA markers were not able to identify population structure (McLean et al. 1999), but highly polymorphic
microsatellite loci (Kaukinen et al. 2004) provided more resolution to define management units (Beacham et al.
2005). In Alaska, analyses with these same microsatellite loci identified two large population groupings, one to the
north encompassing Cook Inlet, Prince William Sound, and Yakutat Forelands and one in the south from populations
within the Alexander Archipelago south to Behm Canal (Flannery et al. 2013). This regional separation follows a
hierarchical island model (Slatkin and Voelm 1991) and did not show isolation-by-distance within each region. The
division of the regions and the lack of IBD within region may be in part due to larval dispersal associated with the
counter-clockwise Alaska gyre dispersing larvae from the Yakutat Forelands to the east; the southern populations
within the Alexander Archipelago would not experience the same transport. Restriction-site associated DNA
sequencing (RADseq) was applied to eulachon collections from 12 populations spanning from the Columbia River
to the Northern Gulf of Alaska (Candy *et al.* 2015). Although sample sizes for the RADseq project were smaller
than those previously analyzed by microsatellites, this study provided strong evidence for three genetic clusters (i.e.,
Northern Gulf of Alaska, Southeast Alaska-BC, and Fraser-Columbia), and markers identified as putative outliers
or having high F_{ST} were identified to be used for future studies. High F_{ST} markers can provide increased
discrimination power for genetic stock identification (GSI; Ackerman et al. 2011; Hess et al. 2011). The combination
of highly resolving markers and high sample sizes per population provides the best power for resolving a species
with such low structure. However, such studies would benefit from having multi-year collections to ensure allele
frequencies are stable over time (i.e., multiple brood cycles) and not highly impacted by sweepstakes reproductive
success (Hedgecock and Pudovkin 2011) and annual variation within rivers (Waples 1998).

Assigning unknown origin fish by GSI and mixed-stock analysis (MSA) to population or region of origin
is useful for interpreting at-sea ecology during surveys or bycatch interceptions, as little is known as to how eulachon
mix among populations at-sea, or where populations are during different seasons. Knowing where specific DUs
exist and are intercepted as bycatch may help reduce impacts on at-risk populations. Eulachon MSA has been
conducted on eulachon bycatch from fisheries off the west coast of British Columbia since the early 2000s using the
microsatellite panel (Beacham et al. 2005). Increased knowledge regarding eulachon at-sea ecology and behaviour,
as well as the ability to identify natal origins of intercepted fish have both been highlighted as priority research needs

in Canada (Schweigert et al. 2012), on the US West Coast (NMFS 2017), and in Alaska (Ormseth et al. 2008). The

microsatellite panel used for MSA identified that eulachon bycatch from a Chatham Sound (North Coast of BC)

shrimp trawl fishery in 2001 were predominantly of central mainland and Nass River origin, whereas eulachon

caught in a research survey in Queen Charlotte Sound (Central Coast of BC) were from various regions coastwide,

and those caught in a survey along West Coast Vancouver Island were predominantly of Columbia and Fraser origin

(Beacham et al. 2005). Tests of samples of spawning eulachon in river mouths (i.e., genetic baseline collections)

within central and southeast Alaska also have demonstrated the potential for assignment within the northern and

southern groupings (Flannery et al. 2013). Although ultimately the resolution ability for GSI depends on the level

of differentiation that exists within the species, advances in genotyping technology makes it possible to develop an

improved baseline with highly resolving markers, large sample sizes, and multiple sampling years, providing more

accurate estimates for MSA and GSI.

Low-density (i.e., ~500 markers) single nucleotide polymorphism (SNP) marker panels genotyped using

high-throughput sequencing of amplicons (also known as GTseq) are routinely applied for GSI and parentage-based

tagging (PBT) in order to genotype thousands of individuals at low costs, particularly in the salmonid fishes. For

example, genetic baselines of Coho Salmon *Oncorhynchus kisutch* are currently being used for MSA in BC fisheries

in combination with PBT to assign unknown origin individuals back to genotyped parents from hatchery broodstock

(Beacham et al. 2019). GTseq is a cost-effective solution for questions that do not require high-density markers, but

that need to be applied to thousands of individuals (Campbell et al. 2015). Several aspects of SNPs are advantageous

compared to microsatellite markers, which may include automation, cost, portability of methods, and scalability

(Hauser et al. 2011). Further, due to the fewer number of alleles per locus compared with microsatellite loci, fewer

individuals are needed in order to generate allele frequencies for baseline samples (Beacham et al. 2011).

The present study aims to develop an approximately 500 marker SNP panel using a contig-level genome

assembly (Sutherland et al. in prep) and the top discriminating markers from the previous coastwide RADseq study

(Candy et al. 2015). The developed SNP panel was applied to tissue archives and new collections to generate a

representative baseline encompassing 1,989 individuals from 14 populations with at least 35 individuals per

population from Northern California to Central Alaska. This will allow us to evaluate any improvements in power

to differentiate between populations and regional groups relative to the previously-used microsatellite baseline. The

overarching goal of the SNP panel and extended eulachon genetic baseline is to ultimately use the improved baseline

to improve our understanding of at-sea ecology of eulachon via GSI and MSA.

136   **METHODS**

Amplicon panel design

Markers were obtained from a RADseq study of eulachon populations (n = 12 populations) from the Columbia River

(Washington, USA) to the Northern Gulf of Alaska (Candy et al. 2015). These 4,104 RAD loci were each 90 bp in

length and were aligned to the eulachon reference genome (Sutherland et al. in prep) using BLAST (Altschul et al.

1990) with a minimum e-value cutoff of 1e-20. Only the loci that aligned to the reference genome two or fewer

times were retained in order to reduce non-specific hits and markers present in repetitive regions of the genome. The

top hit as evaluated by bitscore, e-value, and percent identity per locus was used to retain a single representative hit

per locus for downstream analysis.

A total of 200 bp in each direction flanking the first SNP in the marker was extracted from the reference

genome using a custom pipeline (see Data Accessibility: *fasta_SNP_extract*). If the start or end of a contig was less

than 200 bp from the SNP, then a 400 bp segment was obtained from the start or end of the contig, respectively. The

segment of interest was extracted from the contig using *bedtools* (Quinlan and Hall 2010). The variant locus was

marked with both alleles within the output amplicon fasta file. All putatively adaptive markers from Candy et al.

(2015; n = 193) that passed quality filters were retained for the design (n = 181 putatively adaptive markers).

Putatively neutral markers with the highest overall F_{ST} were added until a total of 600 markers were present in the

panel (n = 419 neutral markers). The panel was then submitted for design through the AgriSeq panel design pipeline

(Thermo Fisher).

Sequencing and variant calling

Eulachon samples in the tissue archive at the Molecular Genetics Laboratory (MGL) at Pacific Biological Station

(Fisheries and Oceans Canada) or provided by collaborators (Table 1; see *Acknowledgements*) were amplified using

the designed panel as per manufacturers' instructions.

DNA was extracted from tissues using a variety of methods, including chelex extraction, a modified version

of the chelex extraction (i.e. PBT chelex) that includes an additional overnight incubation (with Proteinase-K, chelex

beads and UltraPure water) and no high temperature thermal cycling step, the Wizard SV 96 Genomic DNA

Purification System (Promega), DNeasy (QIAGEN), and BioSprint (QIAGEN). Highest genotyping rates were

observed using DNeasy or BioSprint methods, and so when possible, these were used for additional extractions.

Individuals were barcoded and amplified using the AgriSeq panel protocol and the eulachon v.1.0 primers

(Additional File S1) as per manufacturers instructions, using the 768 barcodes available (Thermo Fisher), as

previously described (Beacham et al. 2017). Individuals were multiplexed in batches of 768 and sequenced on each

Ion Torrent PI chip (Thermo Fisher). Sequenced samples were de-multiplexed and variants called using a hotspots

file (Additional File S2) in the Torrent Suite software (TS v.5.10.1; *variantCaller* v.5.6.0.4; Thermo Fisher). Called

variants were then imported into the MGL genotype database management system.

Data filtering

Data filtering was conducted sequentially, where first, samples were removed due to missing genotypes (i.e., retain

individuals with less than 50% missing data, that is, being genotyped at >259 amplicons). The cutoff of >259

amplicons was chosen as 50% of the total ~518 amplicons that typically made it through quality control filters (note:

this number changed slightly depending on the samples included in the analysis). Second, populations with too few

samples were removed (i.e., retain when population has ≥ 20 individuals) for initial evaluation of population

structure. Third, amplicons with excess heterozygosity (≥ 0.5), or those missing in too many individuals ($\geq 50\%$

samples) were removed from the data. Markers were previously screened for deviations from Hardy-Weinberg

equilibrium (Candy et al. 2015). The filtered baseline database was date-stamped and converted to genepop format

for downstream genetic analyses.

Locations of baseline populations were plotted on a map in R (R Core Team 2019) using ggplot2 (Wickham

2016) and ggrepel (Slowikowski 2019) based on the GPS coordinates at the river mouth of baseline sites.

Population differentiation analysis

The baseline genotypes for all individuals were read into R using adegenet (Jombart 2008) using a custom pipeline

(see Data Accessibility; *simple_pop_stats*). Pairwise F_{ST} (Weir and Cockerham 1984) was calculated for all

populations using hierfstat (Goudet 2005). A neighbour-joining tree using the *edwards.dist* distance metric (Cavalli-

Sforza and Edwards 1967) was generated using the aboot function of *poppr* (Kamvar et al. 2014) with 10,000

bootstraps. This was exported in tree format and input to FigTree v1.4.4 for data visualization (Rambaut 2019).

Isolation-by-distance (IBD) was evaluated by finding an approximate distance between all recorded GPS

coordinates for all collection sites using the *distm* function of geosphere (Hijmans 2019) and custom scripts (Data

Accessibility; *simple_pop_stats*). Pairwise F_{ST} and pairwise physical distances (km) were compared to calculate a

linear model best fit line to determine adjusted R^2 values of how well the data fit the model.

To determine the proportion of variance captured within larger groupings observed in the dendrogram,

within populations inside larger groupings (i.e., to determine the amount of variation among collections within a

larger grouping), and the unexplained remaining variation that exists among samples, an Analysis of Molecular

Variance (AMOVA) was calculated using the function *poppr.amova* of poppr that uses the *ade4* package (Dray and

Dufour 2007) using default parameters (e.g. removing loci with more than 5% missing data).

Multivariate statistics

Principal Components Analysis (PCA) was performed by first converting the genind to genlight using the *gi2gl*

function of dartR (Gruber et al. 2018), then conducting a genlight PCA by the glPca function of adegenet, then

plotting with ggplot2 using 95% confidence to draw ellipses around the samples from each grouping using the

function *stat_ellipse* (Wickham 2016). Eigenvalues were plotted, three principal components were retained, and

allele loadings for each PC as characterized by *loadingplot* in adegenet were plotted to identify top loading markers

into PCs. Further, a Discriminant Analysis of Principal Components (DAPC) was performed using adegenet,

retaining 10 PCs and one axis, and variance contributions were plotted. Top loading markers in the DAPC were

characterized.

Relatedness

Inter-individual relatedness within a population was calculated for all populations to compare relative relatedness

values. A genlight object was created using dartR, then converted to Demerelate format (Kraemer and Gerlach 2017)

in order to format using the 'readgenotypedata' function of related (Pew et al. 2015). Subsequently, the coancestry

was calculated within related, implementing *coancestry* (Wang 2011) using the *ritland* (Ritland 1996) and *wang*

(Wang 2002) metrics. Relatedness for these metrics was plotted in R for relatedness within a population.

Microsatellite Data

Microsatellite data were obtained to compare with the baseline data genotyped by the SNP panel. Existing

microsatellite data were obtained from the baseline collection database at MGL (Beacham et al. 2005) using

Microsatellite Manager v.10.3 (Candy et al. 2002). The newly added population from California (Klamath River)

was genotyped using the same genotyping methods as previously described (Beacham et al. 2005). F_{ST}, IBD, and

dendrograms were all calculated as described above.

224   **RESULTS**

Amplicon panel design

Of the total 4,104 SNPs from Candy et al. (2015), 3,957 (96%) were found to have at least one significant alignment

against the eulachon reference genome (Sutherland et al. in prep). From these markers, 3,880 aligned with a single

significant hit, 48 with two hits, and 29 with more than two hits. RAD loci with two or fewer hits were retained for

further development (n = 3,928 RAD loci). These markers were further reduced to a total of 600 SNPs by

preferentially selecting the putatively adaptive loci that passed alignment filters (n = 181 of 193) and high F_{ST} (n =

419) SNPs from Candy et al. (2015). Of these, six putative adaptive and 14 neutral SNPs submitted did not pass

primer design, leaving a total of 580 pairs designed into the eulachon AgriSeq panel (v1.0; Thermo Fisher). This

panel is comprised of primers optimized for Ion Torrent Proton technology, and has not been tested with other

technology. Primer sequences are available in Additional File S1.

Baseline population genotyping and quality control

Four sequencing chips (PI v3; Ion Torrent) were used for direct amplicon sequencing of baseline populations of up

to 768 individuals per chip. Sequencing generated a total of 83.9 M, 58.5 M, 79.7 M, and 48.9 M reads within

amplicons per chip for baseline samples, with an average number of reads per sample of 123 k (median = 102 k;

standard deviation = 112 k), 77 k (med = 33 k; sd = 121 k), 105 k (med = 73 k; sd = 114 k), and 87 k (med = 65 k;

sd = 83 k), respectively. After individual samples with high missing data were removed, a total of 19 populations

with at least 20 individuals per population was retained (Table 1). Markers were identified that had excess

heterozygosity (Figure S1; n = 35 markers) or excess missing data across samples (n = 19). Removing these markers

left a total of 526 amplicons. Five of these markers were monomorphic across the populations genotyped, leaving

521 remaining amplicons, with a single SNP designated per amplicon. This reduced set of markers is the quality

controlled marker set (Additional File S2). After all of the quality control, on average there were 111 samples per

population (sd = 98 samples).

To determine the appropriate minimum sample size threshold per population, all populations with at least
20 individuals were clustered into a dendrogram (Figure S2). The groupings of populations with 35 or fewer
individuals (i.e., Bear River, Falls Creek, Kitimat River, Carroll Creek, and Elwha River) were consistently outside
of the main clusters. Therefore, a threshold of 35 individuals per population was applied (see Figure 1) and any
populations with fewer than 35 individuals were removed from the baseline. This resulted in a total of 1,989
individuals for 14 populations, with an average of 142 individuals per population (sd = 98; min = 51; max = 339).
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Hierarchical population structure
The most divergent population in the dataset was Twentymile River (AK), a river at the upper end of Turnagain
Arm in south-central Alaska (Figure 1). This was indicated by the relatively high genome-wide differentiation when
compared to all other populations (mean F_{ST} = 0.0427; Table 2). For comparison, a population near the middle of
the sampled range, Klinaklini River, compared with all other populations except Twentymile River indicates much
less differentiation (mean F_{ST} = 0.0100). This distinctiveness of Twentymile River also can be observed in a
Principal Components Analysis (PCA) along PC2 (Figure S3). PC2 was separated by numerous markers, although
some contributed more substantially to the division (Figure S4A).

The second largest separation in the data separated populations from the Fraser River and south, grouping
the Fraser, Columbia, and Klamath Rivers (Figure 2). This separation of the northern and southern clusters had high
bootstrap support (> 99.99 %), and was also apparent along PC1 of the PCA (Figure S3). PC1 was also separated
by numerous markers, although ~8 markers showed a high contribution to this separation (Figure S4B). Within the
southern grouping, there was some clustering of Columbia River populations together, but the Cowlitz River
population, the most numerous of the Columbia River collections and entirely sourced from 2002 (Table 1), grouped
into a cluster with Klamath River, and more broadly with the Fraser River (Figure 2), rather than with the other
Columbia River populations (Columbia River, Sandy River). Cowlitz River and Klamath River are grouped closely
together and in 87% of trees group together without the Fraser River. In general these populations were very similar
(e.g., Fraser River vs. Columbia River F_{ST} = 0.0079; Fraser River vs. Klamath River F_{ST} = 0.0021; and Klamath
River vs. Columbia River F_{ST} = 0.0091).

Within the northern grouping, there was a strongly supported cluster including the populations in Johnstone
Strait (Kingcome River and Klinaklini River; 99.8% bootstrap support) and more broadly with Bella Coola (86.84%
bootstrap support; Figure 2). These populations had high genetic similarity with each other (mean F_{ST} = 0.0021;
Table 2). Second, other Central Coast populations Kemano River and Wannock River were grouped together and
were highly similar (93.56% bootstrap; F_{ST} = 0.0043). Although Bella Coola River and Wannock River did not
group in the same cluster as may be expected due to physical proximity, they still showed low differentiation (F_{ST}
= 0.005). North Coast populations Nass River and Skeena River were nearly indistinguishable (87% bootstrap; F_{ST}
= 0.0009). The Nass River was more differentiated from the Klinaklini River, for example (F_{ST} = 0.006). The
Transboundary Region's Unuk River clustered outside of the North Coast and Central Coast groupings, but still
within the larger northern grouping.

Although there are three clear groupings in the data (Gulf of Alaska, northern populations, southern populations), as expected from previous work (Candy et al. 2015), there was evidence for Isolation-by-Distance (IBD) with a linear relationship between pairwise F_{ST} and physical distance (km; adjusted $R^2 = 0.71$). This relationship was structured into three main groupings in the magnitude of F_{ST} differences (Figure 3). Using five tentative reporting units (repunits; i.e., grouping of similar populations) as viewed in the dendrogram and shown by appended regional information in Figure 2, an analysis of molecular variation (AMOVA) was used to view the partitioning of variance within groupings. Although the majority of variation is among individuals within populations (96.34%), the between repunit variation was 2.13%, whereas the between samples within repunit was 1.53% (Table 3).

Annual variance in allele frequencies

Applying the minimum sample size threshold of 35, several populations had sufficient sample sizes to split into different years, maintaining the n = 35 threshold for each population-year combination. Populations with multiple year groups included Bella Coola River, Kingcome River, Skeena River, Klamath River, and Fraser River (Figure S5 and Figure S6). For these populations, there was often close clustering of the different collection years, but not always. For example, close clustering occurred for Skeena River 2010 and 2013 (but not 2001), Klamath River 2013 and 2014, Kingcome River 2002 and 2012, Bella Coola River 1998 and 2017 (but not 2013 or 2018). The Fraser River 2014 collection clustered away from the Columbia River collections, but both the 2019 and especially the 2018 collections were more similar to the Columbia River populations, although bootstrap support values for these positions were low.

Inter-individual relatedness

Relatedness of individuals within a collection was calculated using both the *wang* and *ritland* estimators (Figure S7). Consistencies were noted between the estimators finding numerous outlier related individuals for the Fraser River, Kingcome River, and Wannock River collections. The *ritland* estimator found highly related individuals within the Twentymile River collection (Figure S7A), but this was not consistent with the *wang* statistic, which showed reduced relatedness for this collection (Figure S7B). Notably, the Ritland statistic is more impacted by siblings in the data (Wang 2002).

Comparison with microsatellite data

The microsatellite data were obtained from the most recent database that was originally analyzed in Beacham et al. (2005), with augmentation of several stocks including the newly genotyped Klamath River samples. In total, 13 populations were retained that were common with the SNP baseline and that had greater than 35 individuals per population (Figure S8). On average for the microsatellite data, there were 269 individuals per population (sd = 222; min = 69; max = 736).

319   The general trend of the data was similar between the SNP and microsatellite results, with a large divide

320 between the populations to the south of the Fraser River, inclusive, and the populations to the north of the Fraser

321 River, with Twentymile River as an outgroup (Figure S9). For the microsatellite data, Unuk River was also identified

322 to be an outgroup to the rest of the data. The bootstrap support for the northern and southern general groupings using

323 SNPs (99.99% and 100%, respectively) was higher than that for the microsatellite data (87.92% and 85.81%,

324 respectively). The grouping of the Central Coast and Johnstone Strait (CC-JS) was much less apparent in the

325 microsatellite data, and the Fraser River grouped closely with the Columbia River and south populations in the

326 microsatellite data as well. Interestingly, in both the microsatellite and SNP data, Cowlitz River, Klamath River,

327 and Fraser River grouped together more than any of these grouped with the Columbia River collection (2000).

328 Overall, bootstrap support was higher in the SNP data than the microsatellite, where the SNP data bootstrap values

329 were on average 86.17% (n = 11 values) whereas for the microsatellite data the bootstrap values were on average

330 64.02% (n = 10 values).

331   The microsatellite data also showed evidence for IBD, with in general a lower overall range of F_{ST} values

332 (Figure S10; F_{ST} range: 0 - ~0.012) than that observed in the SNP data (Figure 3; F_{ST} range = 0 – ~0.045). There

333 was also less of a gap between the regions in the microsatellite data. Further, the trend was less linear for the

334 microsatellite data (microsatellite adj. R^2 = 0.48; SNP adj. R^2 = 0.71). Lower F_{ST} values were observed in the

335 microsatellite data, and some populations had negligible differentiation (F_{ST} = 0; Table S1). Similarly, the AMOVA

336 for microsatellite data put into regions as determined in the SNP data and geographically (Table S2) shows that only

337 0.66% of the variation exists between repunits relative to the 2.13% explained in the SNP data. The microsatellite

338 data showed 0.08% of the variation in the data existed between samples within repunit, and 99.26% remaining

339 within samples.

341             **DISCUSSION**

342 Improved genetic techniques may advance our understanding of the ecologically and culturally important eulachon,

343 for example to address questions about reasons for declines in some populations and healthy returns in others that

344 are in nearby rivers. These techniques may further our understanding of eulachon in the ocean, including their

345 distribution, how stocks mix at-sea, what populations end up in by-catch and where this occurs, and what populations

346 are being characterized in scientific surveys. Here we present a SNP amplicon panel and improved eulachon baseline

347 (14 populations, 1,989 individuals) using 521 differentiating markers sourced from a RADseq study (Candy et al.

348 2015) combined with a contig-level genome assembly (Sutherland et al. in prep).

349   The newly developed panel outperforms the existing microsatellite panel (Beacham et al. 2005; Kaukinen

350 et al. 2004) as demonstrated by levels of bootstrap support in dendrograms, improved clustering of populations by

351 geographic region, increased estimates of F_{ST} among populations with fewer samples included in the baseline,

352 increased adherence to the expected isolation-by-distance model, and increased genetic variance captured within

353 and between groupings (i.e., repunits). Putative adaptive markers in the panel have been found to provide higher

levels of differentiation compared to the neutral markers and should provide a better understanding of different
selective pressures occurring over the eulachon range (Candy et al. 2015). The new SNP panel is commercially
available (Thermo Fisher) and primer sequences are provided herein (Additional File S1).

With the SNP marker panel, three main, large-scale groupings are observed, with some sub-structure within
each. This includes the Gulf of Alaska (GoA), southeast Alaska and northern BC, and southern BC through the
contiguous US. The northern grouping may yield further subdivision into the Southeast Alaska/North Coast (SEAK-
NC) and Central Coast/Johnstone Strait (CC-JS) reporting units, although the true separation of these populations
requires additional study through simulations. The southern grouping may yield further subdivision into the Fraser
River (FR) and the contiguous US Pacific Northwest (PNW), which also requires further study through simulations.
Evaluating the effectiveness of more granular resolution within the three identified groups is an important next step
for this work, and will be useful to combine both simulated data with empirical data to determine the optimal
resolution that can be achieved.

Regional grouping, isolation-by-distance, and annual variation
Although in general, a hierarchical island model with IBD explains the variation in the dataset, there were some
exceptions within regions. Within-region IBD was not identified in a recent study of eulachon in Alaska, where the
data were more fit to a hierarchical island model rather than strictly IBD, which was explained as potentially due to
large-scale oceanic currents and larval dispersal (Flannery et al. 2013). In the present study, Bella Coola, Klinaklini
and Kingcome River populations grouped closely, but the Wannock and Kemano River populations from the similar
region grouped separately. The cause of the close grouping of these two separate groups is unknown, but could
reflect genetic variation associated with run timing or aspects of habitat differences in the area (e.g., local
hydrology). However, the genetic differentiation across these groupings is still low (Wannock River vs. Kingcome
River $F_{ST} = 0.0064$; Table 2).

For eulachon, run timing may depend on both water temperature and river discharge rate in local river basins
(Langer et al. 1977; Ricker et al. 1954; Smith and Saalfeld 1955). Run timing variation can indicate the potential for
local adaptation (Beacham *et al.* 2005). Run timing variation occurs across different rivers, generally but not always
along a latitudinal gradient (Hay and McCarter 2000; Moody and Pitcher 2010). Peak run timing of eulachon ranges
from February in the south (e.g., Columbia River; WDFW & ODFW 2005), March in Central and North coasts (e.g.,
Kemano, Bella Coola, Skeena, Nass rivers; Moody 2008) or April (e.g., Kingcome and Klinaklini; Southeast Alaska;
ADF&G 2008; Moody 2008), and May in Alaska (e.g., Central and Western Alaska; Moody 2008). In contrast to
the latitudinal trend, Fraser River (DFO 2020), Sandy River, and Klamath River (Larson and Belchik 1998) have
run timing in late March or early April (reviewed in Moody 2008). However, long-term trends toward earlier return
timing of eulachon have been noted in several rivers (COSEWIC 2011; Gustafson et al., in prep; Moody and Pitcher
2010), and these trends are likely associated with increasing river temperatures or changes in peak river flows. When
environmental conditions are different among locations, and selection acts upon adaptive variants fit to these
conditions, local adaptation is possible if sufficient isolation among populations occurs. Low genetic differentiation

between populations suggests low drift and/or high gene flow, which reduces but does not preclude the potential for local adaptation. Whether the close genetic similarity observed for Wannock and Kemano river populations and similarly for Bella Coola, Kingcome, and Klinaklini rivers, is due to local environmental differences at each location (e.g., river hydrology) is an interesting avenue for future investigation; additional samples from each of the two different groupings should provide more insight as to whether this trend remains.

The present study found consistent grouping of populations separated by year of collection, such as the Klamath River and Kingcome River, and for some but not all collections of Skeena River and Bella Coola River. Nonetheless, for the most part, groupings stayed within their putative reporting unit, and always within their larger regions (i.e., the three main groupings). This is consistent with other studies finding a greater effect of geography than temporal variation (Beacham et al. 2005). Sampling multiple years is a useful method of reducing the variance inherent in collections across years (Waples 1998), and has been highlighted as a valuable step to evaluate potential for mixed-stock analysis accuracy (Flannery et al. 2013). In addition, with climate change scenarios and expected changes in the distribution of species, it will be informative to continue collecting baseline samples in future years to ensure trends remain consistent. It is interesting to note that there may be multiple runs per river, having peak spawning time at different dates, as has been observed in the Nass River, with an initial run arriving in early to mid March, and a second run arriving in early April (COSEWIC 2013; Langer et al. 1977; Noble et al. 2012), as well as the Kingcome River (Gustafson et al. 2010), the Fraser River (LFFA 2015), the Elwha River (Gustafson 2016), and others throughout the range. If this hidden variation within a river is not included, for example in metadata of collections, it could lead to variance occurring across sampling years if there is only a single sampling event per year.

Genetics and current management groupings

Gustafson et al. (2012) identified one Distinct Population Segment (DPS) of eulachon in the California Current. DPSs to the north of the Skeena River were not identified, as the Status Review teams' mandate was to identify a DPS of eulachon that contained the petitioned populations of eulachon from the states of California, Oregon, and Washington, and not to identify DPSs coastwide. Gustafson et al. (2012) suggest the strong ecological and environmental break that occurs between the Alaska and California Currents provide support for discreteness of eulachon within the California Current.

The three Designatable Units (DUs) within Canada (COSEWIC 2011) correspond well with genetic structure observed here, although it is not entirely clear where the separation point exists between the Nass/Skeena DU and the Central Coast, either around Bella Coola or Johnstone Strait. The Nass and Skeena River populations are very similar genetically to the Kemano and Wannock River populations, suggesting these should be collectively considered as one grouping. Whether the Kingcome, Bella Coola, and Klinaklini rivers are distinct enough from these other Central and North Coast populations for sufficient resolution in mixed-stock analysis is an important next step to this work. The present results are in agreement with previous results suggesting high gene flow between the Nass River and Skeena River populations (see COSEWIC 2013). Additional baseline samples in future years

426 should continue to help resolve the groupings and monitor for any changes. These will be continued to be added to
427 the existing baseline, as multiple year collections are known to reduce the potential for single year sampling biases.
428

CONCLUSIONS

Based on genetic evidence, the current baseline with populations from Klamath River in California through coastal
British Columbia and north to Twentymile River in Alaska is tentatively grouped into five reporting units. The
current SNP panel outperformed the microsatellite panel and will be applied to mixed-stock analysis. Low genetic
differentiation was observed overall compared to other anadromous species such as salmonids, although an
isolation-by-distance model best explained the structure observed here. Annual differences in collections were
characterized, and although in general collections clustered together regardless of year, some variance existed,
indicating the value of having these multiple year collections in the data. Several clustering trends remain
unexplained, but may be related to run timing or hydrological differences of the rivers. Although improvements are
expected as the baseline continues to grow, the current baseline is a foundation that will be used for subsequent
mixed-stock analysis. Important next steps will involve simulating and testing mixed-stock samples to determine
the reliability of the five reporting units proposed here from the population structure analysis.
441

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DATA ACCESSIBILITY

The raw data for this analysis is available on the NCBI Short Read Archive (SRA) under BioProject Accession

PRJNA635905 within BioSamples SAMN15057088-SAMN15060309. The original RAD-seq data are available

within Candy et al. (2015). The eulachon ampliseq panel can be ordered using the following catalogue number

from Thermo Fisher: SKU A44467 AgriSeq Custom Panel –DFO_EULACHON20180627.

The analytical pipelines applied in this work are all available on GitHub, including:

Extend RAD marker using genome: https://github.com/bensutherland/fasta_SNP_extraction

Analyze population genetic data: https://github.com/bensutherland/simple_pop_stats

SUPPLEMENTAL INFORMATION

Additional File S1. Primers for the eulachon panel.

Additional File S2. Hotspot file outlining variants and positions.

Table S1. Genetic differentiation as measured by Weir-Cockerham F_{ST} between collections in the microsatellite baseline.

Table S2. AMOVA results for reporting units and collections using data from the microsatellite baseline.

Table S3. Run timing differences for eulachon populations relevant to the study from the literature.

Figure S1. Observed heterozygosity per marker over all samples in the study.

Figure S2. Dendrogram for all populations in the SNP baseline including populations with fewer than 35 individuals.

Figure S3. (A) Principle Components Analysis of the SNP baseline showing PC1 and PC2; and (B) eigenvalues of the different PCs.

Figure S4. PC loading values of each marker for the top three PCs.

Figure S5. Sample size per collection in the SNP baseline separated by year including only collections with more than 35 individuals.

Figure S6. Dendrogram for all populations in the SNP baseline separated by year including only collections with more than 35 individuals.

Figure S7. Inter-individual relatedness within each collection as calculated by (A) the *wang* statistic; and (B) the *ritland* statistic.

Figure S8. Sample size per population in the microsatellite baseline, only including the populations that are present in the SNP baseline.

Figure S9. Dendrogram for all populations in the microsatellite baseline.

Figure S10. Isolation-by-distance plot for the microsatellite baseline comparing pairwise F_{ST} and pairwise physical distance (km).

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TABLES

Table 1. Populations and years included in the study retained after quality control of baseline shown within larger
groupings (i.e., repunits) and approximate location (GPS). The number of samples from each year as well as the
total number of samples for the collection are shown (n). Collections are also shown whether each was in the
original marker discovery (i.e., Candy et al. 2015).
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Repunit	Collection	Sampling Year (n)	Tot. (n)	Approx. Location (GPS)	Incl. in marker discovery? (n)
GoA	Twentymile R.	2001 (99)	99	60.8454, -148.9854	2001 (32)
SEAK-NC	Unuk R.	2011 (14)	51	56.0543, -131.0251	-
		2012 (37)			-
SEAK-NC	Nass R.	2008 (4)	96	54.9769, -129.8894	2008 (41)
		2013 (92)			-
SEAK-NC	Skeena R.	2001 (40)	341	54.1375, -130.0944	-
		2010 (95)			2010 (33)
		2013 (186)			-
		2019 (20)			-
SEAK-NC	Kemano R.	-	95	53.4839, -128.1247	2001 (42)
		2013 (95)			-
SEAK-NC	Wannock R.	2015 (93)	93	51.6789, -127.2506	-
CC-JS	Bella Coola R.	1998 (78)	241	52.3900, -126.7775	-
		2003 (19)			2003 (33)
		2013 (40)			-
		2017 (45)			-
		2018 (59)			-
CC-JS	Klinaklini R.	2002 (87)	87	51.0922, -125.6261	2002 (41)
CC-JS	Kingcome R.	1999 (7)	127	50.9500, -126.2000	-
		2002 (66)			2002 (36)
		2012 (50)			-
		2013 (4)			-
FR	Fraser R.	2009 (30)	339	49.1153, -123.1835	2009 (40)
		2014 (40)			-
		2018 (188)			-
		2019 (81)			
PNW	Columbia R.	2000 (63)	63	46.2293, -123.6446	-
		-			2011-12 (22)
PNW	Cowlitz R.	2002 (181)	181	46.0982, -122.9090	2002 (37)
PNW	Sandy R.	2014 (53)	53	45.5645, -122.3968	-
PNW	Klamath R.	2012 (34)	123	41.5271, -124.0451	-
		2013 (44)			-
		2014 (45)			-

657 **Table 2.** Pairwise genetic differentiation (Weir-Cockerham F_{ST}) between all populations in the filtered amplicon
658 panel baseline. Shading is used to show increasing F_{ST} values.
659

	BEL	COL	COW	FRA	KEM	KIN	KLA	KLI	NAS	SAN	SKE	TWE	UNU	WAN
BEL	NA	0.0189	0.0150	0.0129	0.0041	0.0024	0.0155	0.0026	0.0019	0.0143	0.0016	0.0410	0.0021	0.0050
COL		NA	0.0073	0.0079	0.0254	0.0193	0.0091	0.0189	0.0224	0.0070	0.0217	0.0443	0.0225	0.0230
COW			NA	0.0021	0.0222	0.0144	0.0009	0.0176	0.0175	0.0008	0.0157	0.0417	0.0154	0.0235
FRA				NA	0.0211	0.0114	0.0021	0.0148	0.0157	0.0021	0.0150	0.0449	0.0151	0.0217
KEM					NA	0.0082	0.0230	0.0083	0.0039	0.0218	0.0033	0.0433	0.0049	0.0043
KIN						NA	0.0148	0.0014	0.0052	0.0146	0.0043	0.0443	0.0060	0.0064
KLA							NA	0.0173	0.0163	0.0018	0.0161	0.0418	0.0149	0.0248
KLI								NA	0.0061	0.0160	0.0052	0.0450	0.0059	0.0062
NAS									NA	0.0172	0.0009	0.0406	0.0016	0.0058
SAN										NA	0.0161	0.0415	0.0151	0.0212
SKE											NA	0.0428	0.0010	0.0061
TWE												NA	0.0395	0.0442
UNU													NA	0.0076
WAN														NA

BEL=Bella Coola; COL=Columbia; COW=Cowlitz; FRA=Fraser; KEM=Kemano; KIN=Kingcome;
KLA=Klamath; KLI=Klinaklini; NAS=Nass; SAN=Sandy; SKE=Skeena; TWE=Twentymile; UNU=Unuk;
WAN=Wannock

Table 3. Analysis of molecular variance (AMOVA) results showing sources of variation within the amplicon
panel baseline, using the filtered baseline (i.e., greater than 35 individuals per collection, grouped by repunit as per
Table 1).
668

Source of variation	Degrees of freedom	Sums of squares	Variance components (sigma)	Percentage of variation
Between repunit	4	2268.21	1.0176	2.13
Between samples within repunit	9	1162.54	0.7301	1.53
Within samples	1975	90809.00	45.9792	96.34
Total	1988	94239.74	47.7269	100.00

FIGURES

(A)

(B)

Figure 1. Locations (A) and sample sizes (B) for all collections with more than 35 individuals per collection included in the filtered baseline.

678
679 **Figure 2.** Genetic similarity among populations in the filtered baseline as shown within a neighbour-joining tree
680 (Cavelli-Sforza and Edwards chord distance) rooted with Twentymile River. Branches are coloured by general
681 grouping as shown in Table 1 (light blue = Gulf of Alaska; dark blue = Southeast Alaska, North Coast, and
682 Central Coast; pink = additional Central Coast and Johnstone Strait populations; red = Fraser River; green = US
683 southern populations).

684
685 **Figure 3**. Isolation-by-distance (IBD) as estimated by comparing the approximate physical distance between
686 collection sites and pairwise F_{ST} values using the SNP baseline. Each point represents a pair of values for each
687 single pairwise comparison.