

1 **On the origins of phenotypic parallelism in benthic and limnetic**
2 **stickleback**

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12 **Abstract**

13 Rapid evolution of similar phenotypes in similar environments, giving rise to *in situ* parallel
14 adaptation, is an important hallmark of ecological speciation. However, what appears to be *in*
15 *situ* adaptation can also arise by dispersal of divergent lineages from elsewhere. We test
16 whether two contrasting phenotypes repeatedly evolved in parallel, or have a single origin, in
17 an archetypal example of ecological adaptive radiation: benthic-limnetic three-spined
18 stickleback (*Gasterosteus aculeatus*) across species-pair and solitary lakes in British
19 Columbia. We identify two genomic clusters across freshwater populations, which differ in
20 benthic-limnetic divergent phenotypic traits and separate benthic from limnetic individuals in
21 species pair lakes. Phylogenetic reconstruction and niche evolution modelling both suggest a
22 single evolutionary origin for each of these clusters. We detected strong phylogenetic signal
23 in benthic-limnetic divergent traits, suggesting they are ancestrally retained. Accounting for
24 ancestral state retention, we identify local adaptation of body armour due to the presence of
25 an intraguild predator, the sculpin (*Cottus asper*) and environmental effects of lake depth and
26 pH on body size. Taken together, our results imply a predominant role for retention of
27 ancestral characteristics in driving trait distribution, with further selection imposed on some
28 traits by environmental factors.

29 **Introduction**

30 Parallel occurrence of adaptive phenotypes across similar but geographically separate
31 environments has long fascinated evolutionary biologists. There are two main mechanisms
32 which can explain such a pattern. First, novel adaptive phenotypes may evolve rapidly and
33 repeatedly in response to new ecological opportunity i.e. parallel ecological speciation [1].
34 Alternatively, an adaptive phenotype may arise in a single location and disperse into and or
35 persist only in suitable environments [2-4]. Although these two mechanisms result in the
36 same pattern, they reflect extremely different evolutionary histories: multiple evolutionary
37 origins of the same phenotype vs a single origin. It is therefore necessary to determine which
38 evolutionary history is responsible for apparent parallelism if we are to understand it. There
39 are many definitions for parallel and convergent evolution [5, 6], but here we focus on
40 whether similar phenotypic adaptations share an ancestral genetic basis.

41 Parallel ecological speciation may involve multiple *de novo* mutations, each of which may
42 lead to a similar phenotype but by a slightly different mechanism. In this instance it is easy to
43 conclude multiple independent origins. However, evolution is not linear but often reticulated,
44 and, in many cases, parallel ecological adaptation may involve repeated reuse of long-
45 standing genetic variation i.e. the same, potentially ancient mutation can be introduced to
46 multiple independent populations via admixture [7]. In this case, parallel populations may be
47 young, and have multiple origins, but the mutations responsible for adaptation are shared and
48 may be much older. This scenario is extremely difficult to differentiate from a scenario in
49 which parallel populations themselves have a single origin [8], but it is critical that we
50 attempt to do so in order to understand the underlying processes that shape evolution.

51 The benthic-limnetic axis of stickleback in British Columbia ('BC'), Canada, is an archetypal
52 example of ecological divergence and speciation [9-14]. It separates bottom-dwelling, benthic

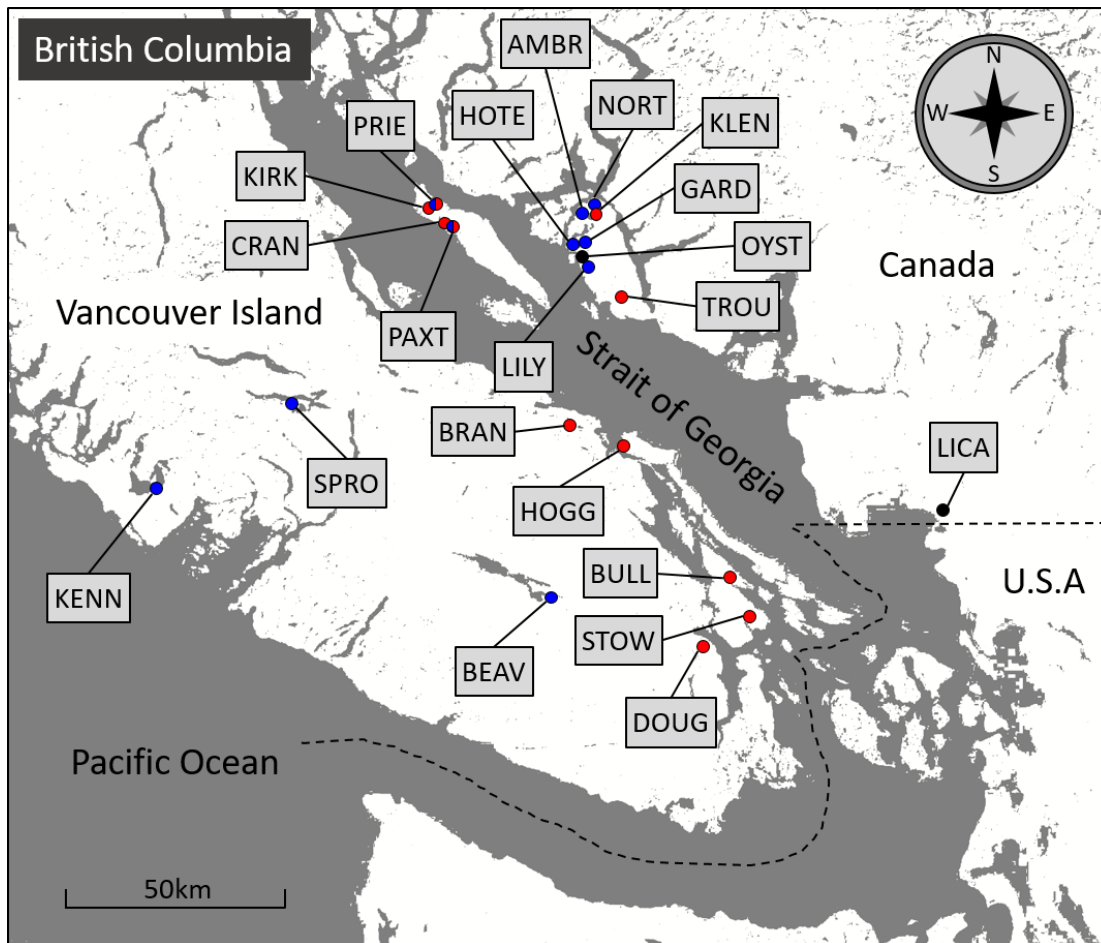
53 individuals, which feed predominantly on macroinvertebrates, from pelagic fish, feeding
54 mostly on zooplankton [15-17]. These two freshwater ecotypes are characterised by heritable
55 differences in body size, shape, trophic morphology and body armour, which confer fitness
56 advantages in their corresponding habitats [18-20]. In BC, stickleback occur both as
57 sympatric benthic-limnetic species pairs and solitary populations that possess phenotypes
58 along the benthic-limnetic axis [9, 21-24]. Previous work has identified patterns of
59 parallelism in adaptive genomic divergence across benthic-limnetic species pairs, but closer
60 genetic affinity within lakes at neutral markers [21, 22, 25]. This work has tentatively led to
61 the conclusion that benthic and limnetic phenotypes evolved repeatedly and independently in
62 multiple lakes [21, 25, 26]. However, gene flow occurs to some extent in all benthic-limnetic
63 species pairs [27, 28], and even low levels of gene flow quickly erode genetic differences at
64 neutral loci, making it impossible to separate patterns of recent *in situ* ecological speciation
65 from those derived from secondary contact of much older independent lineages [29]. Little
66 investigation has so far been conducted beyond the species pairs, which coexist in only a
67 handful of lakes [30], but see [31]. Populations in solitary lakes have far less opportunity for
68 gene flow and thus will likely give a more reliable estimate of the evolutionary history of
69 benthic and limnetic ecotypes.

70 We investigate whether benthic-limnetic divergence in BC stickleback likely has a single or
71 multiple evolutionary origins. We first characterise genomic and phenotypic divergence
72 across populations and show that all freshwater individuals fall within one of two genomic
73 clusters, one of which exhibits a more benthic phenotype, and the other, a more limnetic
74 phenotype. We construct a maximum likelihood phylogeny using a stringently filtered
75 dataset, removing all known QTL in stickleback, and test for phylogenetic signal in
76 ecologically relevant phenotypic traits. We construct a microevolutionary adaptive landscape
77 for the BC radiation using recently available niche modelling techniques [32] to identify the

78 best fitting model of benthic-limnetic trait evolution. Finally, accounting for any phylogenetic
79 signal, we test for relationships between phenotype and environment to detect signals of true
80 ecological adaptation.

81 **Results**

82 We collected stickleback, and environmental parameters from 21 lakes surrounding the Strait
83 of Georgia, BC (Figure 1), including two species pair lakes, two coastal locations
84 (representing putative marine ancestors) and 17 solitary freshwater lakes (Table S1). We
85 collected phenotypic data for key benthic-limnetic divergent traits (methods) for
86 approximately 32 individuals (mean = 31.5, SE = 2.4) and generated RAD-seq genomic data
87 [13], for approximately 16 individuals (mean = 15.9, SE = 0.9), from each lake, 333
88 individuals in total. This resulted in a master genomic dataset of 12,756 SNPs, which was
89 subject to further filtering for some analyses (Table 1).



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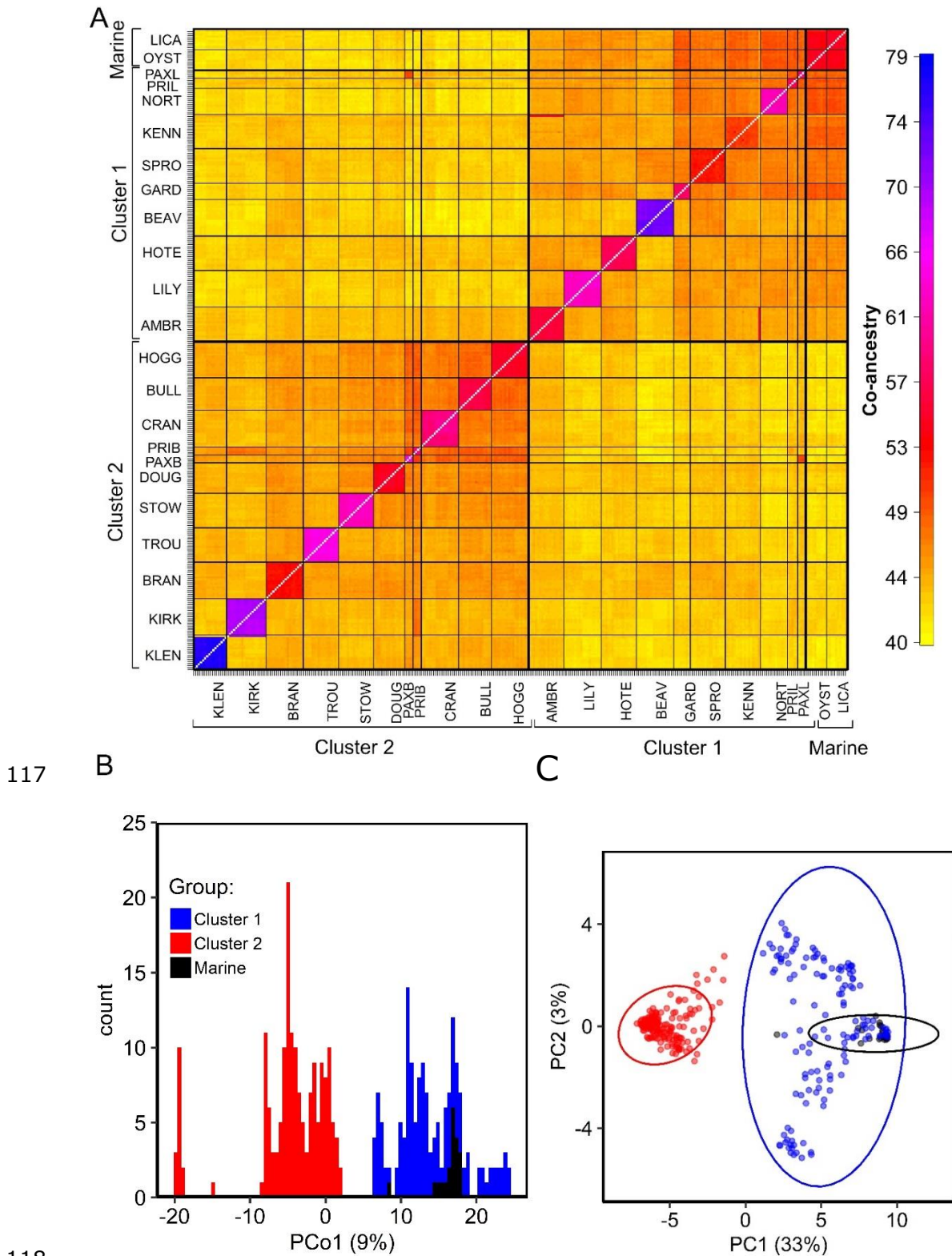
91 **Figure 1.** Map of sampling locations in British Columbia. Sample sites are indicated by
92 circles and their associated lake ID. Black circles indicate marine populations, blue circles
93 indicate populations in cluster 1 of our genomic analyses and red circles, cluster 2. Red and
94 blue semi-circles indicate species-pair populations containing individuals from both clusters 1
95 and 2. The dashed line represents the border between Canada and the USA.

96 *Genomic divergence*

97 We used two methods to quantify clustering within the genomic data. Firstly, a co-ancestry
98 matrix in fineRADstructure [33] (dataset 1, 12,756 SNPs) identified two genomic clusters
99 across all populations (Figure 2A), one incorporating the marine populations and
100 approximately half of the freshwater populations (cluster 1), and the other comprising the rest
101 of the freshwater lakes (cluster 2). Although marine populations formed part of cluster 1, they
102 are considered separately here and in all further analyses because their presence in shallow

103 coastal areas is transient and they likely represent the ancestral phenotypic state of all
104 freshwater populations [7]. Secondly, we conducted a principal coordinate analysis (PCoA,
105 dataset 2, 9668 SNPs). The same two broad genomic clusters (1 and 2) were identified by
106 PCoA analysis, separating along PCo1 (9% of total variation, Figure 2B).

107 To further investigate the genomic properties of cluster 1 and 2 differentiation, we conducted
108 a linkage disequilibrium network analysis (LDna) using the LDna R package (dataset 1,
109 12,756 SNPs). The resulting LD network contained twelve linkage clusters, each of which is
110 likely a signature of a different evolutionary phenomenon [34]. Principal component analysis
111 (PCA) on the SNPs from each cluster revealed a group of 60 SNPs, spread across 17 of the
112 21 chromosomes in the stickleback genome, associated with cluster 1 – cluster 2 separation.
113 Of these 60 SNPs, 28 fell directly within genes (Table S2). Most other LD clusters only
114 separated single populations from all others, likely reflecting local patterns of selection and
115 drift and none of the LD clusters separated marine and freshwater adapted populations
116 (Figures S1 and S2).

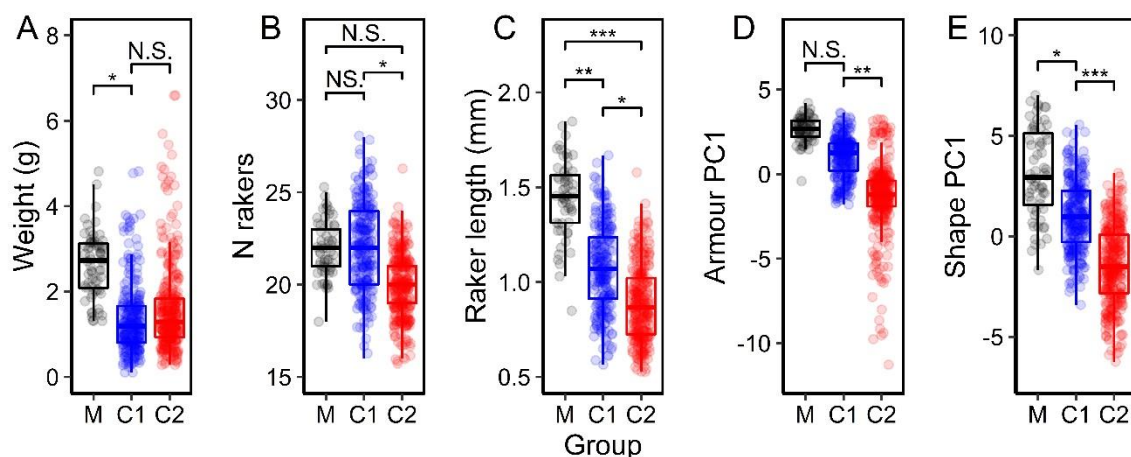


119 **Figure 2.** Genetic structure across BC stickleback populations. (A) Co-ancestry matrix of BC
 120 stickleback populations, using fineRADstructure. Thin black lines separate populations and
 121 thick black lines separate the broader genetic clusters. (B) Distribution of BC stickleback
 122 along the first principal coordinate of a genomic PCoA (dataset 2, 9668 SNPs). (C)

123 Distribution of BC individuals in a PCA of 60 linked SNPs comprising linkage cluster 10,
124 identified by LDna.

125 *Phenotypic divergence*

126 To determine whether the genomic clusters differed phenotypically, we analysed differences
127 in group means for important benthic-limnetic divergent phenotypic traits: weight, gill raker
128 length and number, armour PC1 (increasing size of all armour variables and increasing lateral
129 plate number, explaining 70.4% of body armour variation [methods]) and shape PC1
130 (describing shape changes associated with benthic and limnetic habitats, such as eye size,
131 body depth and mouth length, explaining 23.2% of body shape variation [35, 36]). There
132 were differences in phenotype between the three groups for all phenotypic traits (Table S3).
133 For most traits, clusters 1 and 2 differed from marine fish, and for all traits except for body
134 weight, clusters 1 and 2 differed from each other (Table S4, Figure 3A-E). Cluster 1 had a
135 typically limnetic phenotype [35, 36] with a smaller size, longer, more numerous gill rakers,
136 more body armour, a larger pelvis relative to spine length, and a more streamlined, slender
137 body shape than cluster 2, which had a much more benthic phenotype (Figure 3A-E).



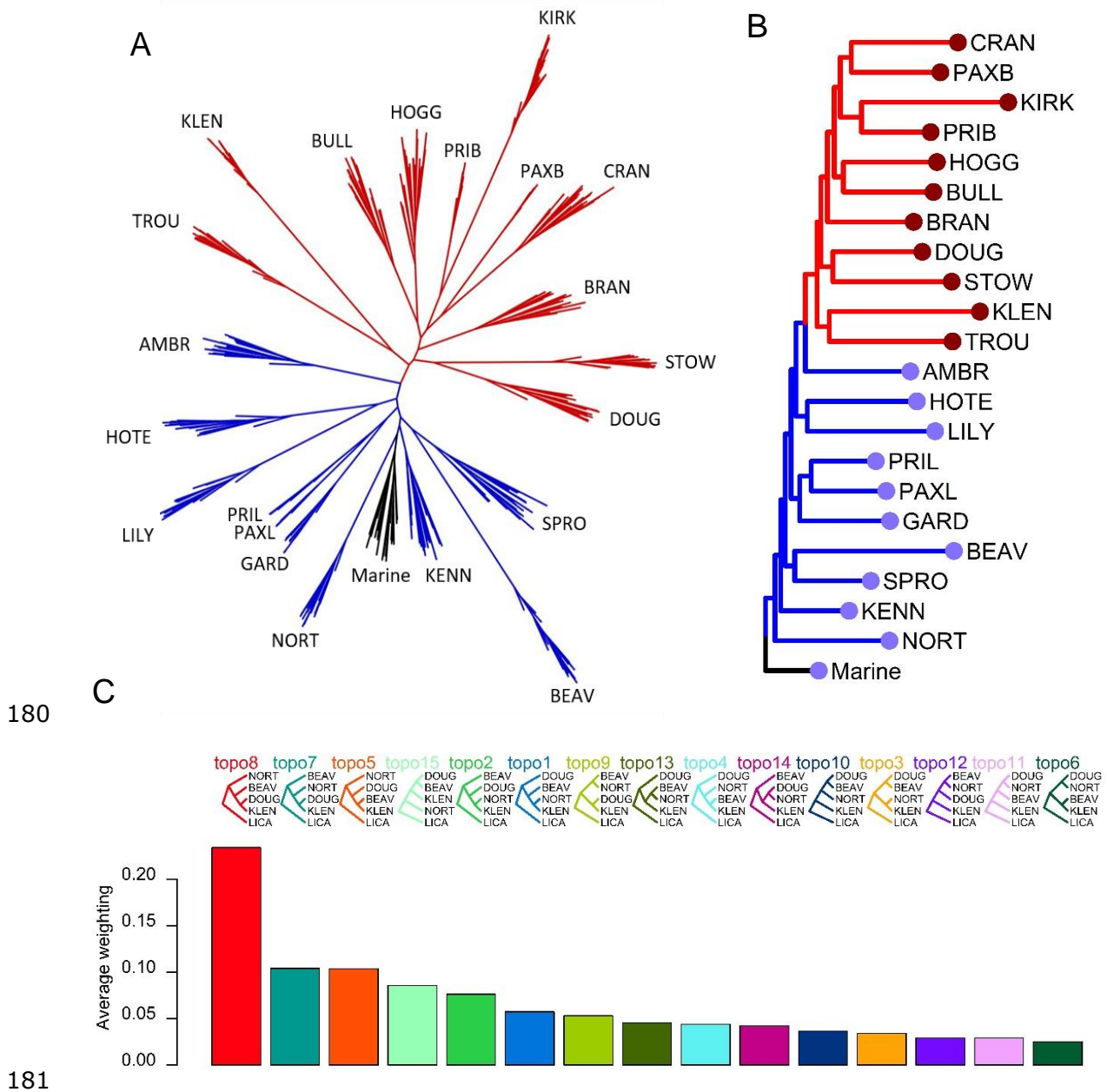
138 **Figure 3.** (A) – (E) Phenotypic differences between marine fish and two freshwater genetic
139 clusters. Circles represent individuals. Abbreviations: M – marine, C1 – cluster 1, C2 –
140 cluster 2. Brackets and asterisks indicate significance thresholds of post-hoc estimated
141 marginal means tests between groups: NS. indicating $p > 0.05$, ** indicating $p < 0.01$ and ***
142 indicating $p < 0.001$. All p values were adjusted for multiple comparisons.

143 *Phylogeny*

144 Phylogenetic reconstruction for population-level genomic data can be notoriously
145 problematic as numerous factors, including ongoing and / or historic gene flow can mask true
146 phylogenetic signal in the data [37, 38]. To minimise bias in our analysis, we first filtered our
147 master dataset (dataset 1, 12,756 SNPs) to remove all known QTL in stickleback (leaving
148 8351 SNPs, see methods), and then filtered for linkage disequilibrium ($R^2 > 0.2$, leaving
149 6215 SNPs, dataset 3). We constructed a maximum likelihood phylogeny for all populations
150 using RAxML. Consistent with the co-ancestry and PCoA analyses, the topology showed
151 clusters 1 (more limnetic phenotype) and 2 (more benthic phenotype) at opposite ends of the
152 tree, with marine fish most closely related to cluster 1 (Figure 4A). The two species-pair
153 lakes both contained limnetic individuals whose closest relatives were in cluster 1 (PAXL
154 and PRIL), and benthic individuals whose closest relatives were in cluster 2 (PAXB and
155 PRIB, Figure 4A).

156 We also performed a topology weighting analysis on a subset of populations selected
157 specifically to test the likelihood that the phenotypes associated with clusters 1 and 2 could
158 have evolved repeatedly *in situ*. Topology weighting is a means by which to quantify
159 relationships between taxa that are not necessarily monophyletic. It determines how support
160 for each possible topology varies across the genome and allows quantification of the overall
161 proportion of the genome which supports each possible tree. This allowed us to identify
162 multiple highly supported phylogenies so that we could determine whether any of those with
163 high support involved a model in which the two clusters arose more than once. It also
164 allowed us to quantify what proportion of the genome supports our most likely topology, and
165 how big the difference is between this and the level of support for the next most likely tree.
166 To do this, we selected two populations from each of two locations approximately 100km
167 apart and separated by the Georgia Strait. Each location contained a cluster 1 and 2

168 population occurring reasonably close together which, under a model based purely on
169 geography, would be predicted to be more closely related. One marine individual was used as
170 an outgroup. The topology with the highest weighting across all 50bp sliding windows
171 (topology 8) was concordant with the maximum likelihood phylogeny, with the two cluster 2
172 populations (DOUG and KLEN) forming a monophyletic clade and each cluster 1 population
173 splitting off earlier, deeper to the root (Figure 4C). The topology with the second highest
174 weightings (topology 7) was also concordant, and just involved a switching of the order in
175 which the cluster 1 populations split from the root. The simple geographical hypothesis, with
176 the two pairs of populations nearest to one another being most closely related (topology 3)
177 received very little support. The highest ranking topology had more than twice the
178 proportional support that the second most likely topology had, suggesting that there is a
179 strong genome-wide signal in favour of the maximum likelihood topology.



182 **Figure 4.** (A) Maximum likelihood phylogeny of 333 BC stickleback. Black indicates
 183 marine; blue, cluster 1 and red, cluster 2. (B) The same phylogeny as (A) with monophyletic
 184 populations collapsed into single tips. Branch colours in (B) denote the same as in (A).
 185 Coloured circles at branch tips represent two independent selection regimes, detected in the
 186 optimal model of niche evolution (R package: SURFACE). In both phylogenies species pairs
 187 are divided into benthic (PAXB, PRIB) and limnetic (PAXL, PRIL) populations. (C) Mean
 188 weightings for all possible topologies for four freshwater populations: two from cluster 1,
 189 NORT and BEAV; and two from cluster 2, DOUG and KLEN, with a single individual from
 190 the marine population LICA as the outgroup.

191 *Phylogenetic signal*

192 If benthic and limnetic phenotypes had resulted from repeated, rapid adaptive divergence,
193 phylogenetic signal (the tendency for more closely related individuals to share phenotypes)
194 would be obscured, and trait distributions would instead mimic the adaptive landscape — i.e.
195 variation in the relevant environmental characteristics. Therefore, we tested a null model that
196 traits would be distributed randomly with respect to phylogeny, and an association of trait
197 distribution with population-level relatedness was taken as evidence that benthic and limnetic
198 niches were conserved from the ancestral state [39-41].

199 We estimated phylogenetic signal at the population level, using mean phenotypic trait data,
200 and collapsing nodes in the phylogeny by population (with the two marine populations
201 grouped into a single node, as they lacked monophyly), using the R package: PhyloSignal.
202 We also tested five simulated traits that had no true association with phylogeny. We
203 identified phylogenetic signal in all five real phenotypic traits: weight, gill raker number, gill
204 raker length, armour PC1 and shape PC1 (p -values < 0.05 , Table 2). None of the five
205 randomly simulated traits showed phylogenetic signal (p -values > 0.05 , Table 2).

206 *Niche evolution modelling*

207 Niche evolution modelling combines phylogenetic information with the distribution of
208 phenotypic traits across the tree to identify the most likely number and location of selection
209 regimes imposed across the whole phylogeny. It also identifies the number of instances of
210 convergence (where the same regime appears multiple times across the tree). If the benthic
211 phenotype had evolved repeatedly and independently across the phylogeny, niche evolution
212 modelling should identify multiple instances of convergence of a benthic selection regime.
213 We performed niche evolution modelling using the R package SURFACE [32]. We ran
214 SURFACE using the same collapsed phylogeny and associated trait data that were used to

215 estimate phylogenetic signal. The best fitting model involved two different selection regimes
216 across the phylogeny (Figure 4B). The first included all marine and cluster 1 populations, and
217 the second, all cluster 2 populations. The best fitting model included no instances of
218 convergence between selection regimes, i.e. each independent regime appeared only once
219 across the phylogeny.

220 *Relationship between phenotype and environment*

221 To test for phenotype – environment relationships, we used linear mixed models, following a
222 phylogenetic generalised least squares (PGLS) approach so that phylogenetic signal could be
223 accounted for and fitted to the population means of phenotypic traits in R. Marine fish were
224 excluded from all phenotype – environment modelling, because of the difficulty of measuring
225 the environment of migratory marine fish. We found that freshwater fish had more body
226 armour in the presence of sculpin (adjusted p -value < 0.05 , Table S5), and fish were heavier
227 in deeper lakes and lakes with a higher pH (adjusted p -values < 0.01 and < 0.05 ,
228 respectively). Lake surface area and calcium concentration did not affect any aspect of
229 phenotype and none of the environmental variables we measured affected the number of gill
230 rakers, the length of gill rakers or shape PC1 (Table S5).

231 **Discussion**

232 The repeated occurrence of similar phenotypes in geographically isolated but similar
233 environments has several possible evolutionary explanations. Perhaps this pattern results
234 from parallel ecological speciation, or maybe similar phenotypes have a single origin and
235 have subsequently become widely dispersed into suitable habitats. It is impossible to separate
236 these different models using only phenotypic data or small numbers of genetic markers, and
237 remains difficult even with genomic data. Nevertheless, it is important that we attempt to do
238 so, because of the consequences for our understanding of evolution. Parallel evolution has
239 been implicated in the global adaptation of stickleback to freshwater [21], but recent analyses
240 suggest that this is unlikely outside of the Eastern Pacific [42]. Furthermore, there are a
241 number of other cases in which conclusions of parallel ecological speciation have been called
242 into question by the confounding possibility of a single evolutionary origin followed by
243 migration and gene flow [29].

244 We investigated the evolutionary origins of divergent phenotypes in a classic model system
245 for adaptive radiation and ecological speciation. We find strong evidence for a monophyletic
246 clade of stickleback with a benthic phenotype distributed across freshwater lakes in the
247 southern Georgia Strait region of BC. The evidence strongly suggests that this clade has a
248 single evolutionary origin, and is derived from a local ancestor with a limnetic phenotype.
249 The benthic clade also encompasses benthic fish from two benthic-limnetic species pairs. Our
250 results are consistent with a single origin for benthic fish in BC rather than repeated
251 independent evolution of the benthic phenotype in multiple lakes. This contradicts the
252 currently favoured model for the evolution of benthic and limnetic stickleback in BC [21, 22,
253 25], highlighting the challenging nature of phylogenetic reconstruction at the population
254 level.

255 Many factors, such as incomplete lineage sorting, hybridization, gene duplication, natural
256 selection and recombination can lead to genealogical discordance in estimations of
257 phylogenetic relationships [43]. Resolving the true relationships between divergent groups
258 can therefore be challenging and require a large number of genetic markers. Much of the
259 current research on benthic and limnetic stickleback in BC has been based only on
260 mitochondrial haplotypes [25] or relatively small SNP sets [21], and has largely been
261 restricted to species pairs. In species pairs, multiple QTL regions are repeatedly responsible
262 for benthic adaptation [21, 44, 45], which is consistent with a single benthic origin, but
263 neutral SNPs imply closer genetic affinity of benthics and limnetics within lakes [21],
264 consistent with multiple independent origins. However, elevated levels of genetic similarity
265 at neutral markers in species pairs would be expected even with low levels of gene flow and
266 thus may not reflect true phylogenetic relationships [27]. Clearly, it is important also to
267 consider relationships in solitary populations, where opportunity for gene flow is greatly
268 reduced. Harer et al. [31] looked at both species pairs and solitary populations and identified
269 considerable genomic parallelism in the former but not the latter. However, they define the
270 benthic-limnetic spectrum in solitary populations solely using lake surface area, which
271 correlates only weakly with phenotype, and does not account for differences in lake depth
272 and the presence of specific predators, which also have a major influence on the availability
273 of benthic and limnetic niches to stickleback [46].

274 We identify a monophyletic lineage within BC stickleback, which has a consistently benthic
275 phenotype when compared with other freshwater or marine populations in BC. Some
276 argument remains among evolutionary biologists about whether pervasive, genome-wide
277 selection, can overwhelm the signal from other markers and obscure tree topologies [43, 47].
278 Sculpin (*Cottus asper*) are an intraguild predator of stickleback, which, when present, select
279 for a more limnetic phenotype [48, 49]. Our results were consistent with this as we identified

280 an effect of sculpin presence on body armour after accounting for phylogenetic signal in our
281 analysis. However, we find the possibility that selection from sculpin obscures the true
282 relationships between populations in our phylogeny unlikely for a number of reasons. Firstly,
283 such a phenomenon is certainly possible in studies using only a small number of markers
284 [50], but with many thousands of unlinked genetic markers, such as in this case, the
285 probability of selection overwhelming the signal from neutral markers is very low [47, 51].
286 Secondly, recent modelling suggests that even with strong selection affecting 10-20% of
287 markers, in most instances, phylogenetic inference remains robust to the effects of selection
288 [52]. Selection from sculpin likely affects less than 2% of the genome [48] and thus is at least
289 an order of magnitude smaller in effect size than would be necessary to obscure the true tree
290 topology in this case. Furthermore, sculpin were present in some lakes containing the benthic
291 clade and not in all lakes containing the limnetic clade. Thus, the genomic groups identified
292 here do not simply mirror the occurrence of this selective agent, but rather represent a deeper
293 set of ancestral relationships.

294 If the benthic phenotype in BC has a single origin, a clear question that requires explanation
295 is how this ecotype spread across BC? Although the lakes containing stickleback in BC are
296 not particularly widespread, they are physically separated by land or ocean, which likely
297 makes dispersal a challenge for freshwater stickleback. We speculate that evidence for a large
298 flood (~500km³ of water) in the Fraser River valley, dated approximately to the end of the
299 Pleistocene and caused by the failure of a large ice dam [53], could provide an explanation.
300 The estimated extent of the flood across the southern Georgia Strait is very similar to the
301 current known distribution of benthic stickleback in BC, raising the tantalising possibility that
302 it may have been responsible for the spread of the benthic lineage of stickleback from a
303 palaeolake in the Fraser Valley, consistent with previous inference about the evolution of
304 Eastern Pacific freshwater stickleback [42].

305 Our investigations have shown that the well-studied benthic-limnetic species pairs should be
306 understood as part of a broader radiation along the benthic-limnetic axis in BC. We highlight
307 the need to consider carefully all possible explanations for the occurrence of parallel
308 phenotypes if we are to achieve a proper understanding of the evolutionary processes that
309 mediate divergence. Stickleback are clearly capable of remarkably rapid ecological
310 adaptation [54-58], but we have shown that the retention of ancestral characteristics can also
311 be important in explaining the distribution of divergent phenotypes. This has significant
312 implications for how we think about the process of evolution and raises the possibility that
313 other model examples of *in situ* ecological adaptation may also result from dispersal rather
314 than convergence.

315 **Declaration of Interests**

316 The authors declare no competing interests.

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326 **Tables**

327 **Table 1**

328 **SNP datasets.** Details of the SNP datasets used in genomic analyses.

Dataset	N lakes	N individuals	N SNPs	LD thinned	Known QTL removed
dataset 1	21	333	12,756	✘	✘
dataset 2	21	333	9668	✓	✘
dataset 3	21	333	6215	✓	✓
dataset 4	5	53	9668	✓	✘

329 N: number, LD thinned: SNPs with $r^2 > 0.2$ removed, Known QTL: SNPs identified as falling
 330 within known QTL regions [59].

331 **Table 2**

332 **Phylogenetic signal in real and simulated phenotypic traits.** Table shows estimates of
 333 phylogenetic signal (Pagel's λ) and their associated p -values. P -values < 0.05 are highlighted
 334 in bold.

Trait	Pagel's λ	p-value
Weight	1.5881	0.0421
Number of gill rakers	2.1535	0.0010
Mean gill raker length	1.6492	0.0032
Armour PC1	0.9667	0.0269
Shape PC1	1.3647	0.0010
Random 1	0.0001	1.0000
Random 2	0.0001	1.0000
Random 3	0.0001	1.0000
Random 4	0.0001	1.0000
Random 5	0.0001	1.0000

335

336 **Methods**

337 *Samples sites and environmental measurements*

338 A total of 21 lakes surrounding the Strait of Georgia, BC, which were likely to vary
339 substantially in the ecological niches they presented to stickleback (because of variation in
340 environmental factors), were selected for sampling (see Table S1 for detailed sample site
341 information and Figure 1 for a map of sampling locations). This included two lakes, Paxton
342 (PAXT) and Priest (PRIE), known to contain benthic-limnetic stickleback species pairs [16,
343 26], and two coastal locations accessible from the sea, Oyster lagoon (OYST) and Little
344 Campbell River (LICA), where marine fish are present during the spring breeding season.

345 The size and depth of a lake largely determine whether both benthic and limnetic habitats are
346 present (in larger deeper lakes) or just benthic (in small, shallow lakes). Therefore, we
347 measured the surface area (km²) using GoogleEarth and collected data on the mean depth (m)
348 of each lake from either HabitatWizard [60] or from data collected in Vamosi [61], with
349 permission. The presence of other fish species can also determine whether both, one or none
350 of those niches are available to stickleback [61]. Many other fish species occur in BC, some
351 of which are predators and/or competitors of stickleback. Cutthroat trout (*Oncorhynchus*
352 *clarkii*) and rainbow trout (*Oncorhynchus mykiss*) are major intraguild predators of
353 stickleback, but both occur in both the littoral and pelagic zones [62, 63] and do not eliminate
354 either niche for stickleback and so are not considered further here. Prickly sculpin (*Cottus*
355 *asper*) are a benthic intraguild predator, and their presence selects for a more limnetic
356 stickleback ecotype [46]. We therefore collected data on the presence/absence of prickly
357 sculpin in all sampling locations from Hutchinson et al. [64], Miller et al. [48], Atkinson [65],
358 Dennenmoser et al. [66] and Vamosi [61] (see Supplementary Material for data sources for
359 each lake).

360 The pH [67] and dissolved calcium concentrations [68] of lake water have previously been
361 associated with external bony armour in stickleback (a trait which varies between benthic and
362 limnetic ecotypes). Therefore we also measured these variables, the former with a calibrated
363 pH meter (Multi 340i, WTW, Weilheim, Germany) and the latter were obtained by collecting
364 two filtered water samples (one acidified with 2% nitric acid, one frozen) from each lake. The
365 dissolved calcium concentration (to the nearest mg/L) was then measured from the water
366 samples at the Division of Agriculture & Environmental Science at the University of
367 Nottingham by inductively coupled plasma mass spectrometry (ICP-MS).

368 *Stickleback sampling*

369 Stickleback were caught using unbaited minnow traps set overnight from the lake shores
370 during spring of 2015 (all stickleback ecotypes move to the shallows during the spring to
371 breed). Samples of between 10 and 63 individuals (See Table S1 for lake specific sample
372 sizes) were taken from each lake and transported to a rental property in aerated lake water for
373 processing. Immediately prior to processing, fish were euthanised with an overdose of
374 tricaine methanesulfonate ('MS222') (400 mg L⁻¹), and killed by destruction of the brain, in
375 accordance with Schedule One of UK Home Office regulations and with the approval of the
376 University of British Columbia Animal Care Committee (UBC animal care certificate A11-
377 0402). Fin clips were immediately taken and stored in 90% Ethanol for later genomic
378 analyses.

379 *Identification of benthic-limnetic divergence*

380 *Phenotypic quantification*

381 Fish sampled from lakes containing species pairs (PAXT and PRIE) were visually classified
382 as benthic or limnetic at the time of capture as well as being later measured for all phenotypic
383 traits.

384 To determine body size, fish were blotted and weighed to the nearest milligram. To assess
385 body shape differences, each stickleback's left side was photographed using a tripod mounted
386 digital SLR camera fitted with a macro lens and macro digital ring flash. Images were scaled,
387 and 13 landmarks were placed on each image using tpsDig, version 2.16 [69]. Landmark data
388 were then exported to MorphoJ, version 1.06d [70]. A Procrustes fit was performed to align
389 specimens by their main axes and remove size and rotation bias. Differences between lakes
390 were identified using a Procrustes ANOVA with lake as the classifier. Allometric variation in
391 body shape was removed by taking the residuals of a multivariate partial least squares
392 regression against log centroid size, and the regression was pooled within lakes because the
393 Procrustes ANOVA indicated differences between group centroids [71]. Regression residuals
394 were exported into R, version 3.5.2 [72], where they were standardised and scaled, and
395 variation in body shape was reduced to a single axis using a principal components analysis
396 (PCA), implemented by singular value decomposition. This principal axis (shape PC1) was
397 used to describe differences in body shape in all further analyses.

398 To assess differences in body armour, fish were first bleached and then stained with alizarin
399 red to highlight external skeletal structures following standard procedure [73]. Fish were then
400 re-photographed as above, images were scaled, and counts of lateral plate number, alongside
401 measurements of standard length, first and second dorsal spine length, longest plate length,
402 pelvis height, pelvis length and pelvic spine length, were taken (continuous elements to the
403 nearest 0.01mm) using ImageJ, version 1.52a [74]. All continuous armour variables (thus
404 excluding plate number, which was independent of body size in our data set) were size-
405 standardized by taking the residuals of a regression against standard length. Body armour
406 variables were highly correlated, thus we used a principal components analysis (PCA) to
407 reduce variation in body armour variables to a single axis: armour PC1. Armour PC1 was
408 used to describe differences in body armour in all further analyses.

409 Finally, the left primary gill arch was extracted from each individual. For each gill arch, the
410 total number of gill rakers were counted, and the mean gill raker length was calculated by
411 taking the mean of the length of the first three rakers on each arch, measured to the nearest
412 micrometre.

413 *Genomic SNP analyses*

414 DNA was isolated from fin tissue using Quiagen Blood and Tissue DNA purification kits.
415 RAD-seq data was generated following Magalhaes et al. [75]. BAM files were produced
416 following Magalhaes et al. [75]. Variants were called from per-individual BAM files to create
417 a single VCF file using the Stacks pipeline [76] in Stacks, version 1.47. The POPULATIONS
418 program in Stacks was run with the following filters: SNPs with a minimum depth of
419 coverage < 3 were removed; SNPs present in $< 50\%$ of individuals within a population were
420 removed; SNPs with a minor allele frequency < 0.05 were removed; and SNPs that were not
421 present in all of the populations were removed. VCFtools, version 0.1.16 [77], was then used
422 to remove sites with mean depth values (over all individuals) < 6 and > 200 , sites with $> 25\%$
423 missing data, sites with a minor allele count over all individuals < 2 and the sex
424 chromosomes. This pipeline produced an overall dataset of 12,756 SNPs for 333 individuals
425 across the 21 lakes (dataset 1). This dataset was then subject to further filtering for some
426 analyses, and detailed information about individual RAD datasets is given in Table 1.

427 *Linkage disequilibrium*

428 Sets of loci that have a tendency to be inherited together, and thus are highly correlated, tend
429 to be affected by the same evolutionary processes and so contain useful information for
430 identifying the characteristics of the processes affecting each set of linked loci e.g. whether
431 divergence is likely linked to small genomic regions e.g. inversions, or is genome wide. To
432 investigate whether any groups of linked loci would distinguish the genomic clusters

433 identified in other genomic analyses we performed a linkage disequilibrium network analysis
434 (LDna) using the LDna package in R. The r^2 linkage disequilibrium matrix was generated
435 using dataset 1 (12,756 SNPs) in Plink version 1.9 [78]. For the extractClusters step of LDna
436 the minimum number of edges was set to 100 and Φ was set to five. SNPs in each LD cluster
437 were extracted from dataset 1 using VCFtools, VCF files were read into R using the vcfR
438 package and principal components analysis (PCA) of the SNPs in each LD cluster was
439 performed using the adegenet [79] package.

440 Many genomic tools, however, rely on the assumption that variants are independent and
441 therefore SNPs in linkage disequilibrium must be removed for such analyses. To that end, we
442 estimated linkage disequilibrium across the genome as a whole by calculating pairwise R^2
443 values in 100kb sliding windows using Plink version 1.9. R^2 values range between 0 (no
444 linkage) and 1 (complete linkage), and therefore a relatively conservative LD threshold was
445 set at $R^2 > 0.2$. Thinning dataset 1 (12,756 SNPs) to unlinked loci resulted in a dataset with
446 9,668 retained SNPs (dataset 2, Table 1).

447 *Genomic patterns*

448 We used fineRADstructure [33] to construct a co-ancestry matrix using the primary SNP set
449 including all 333 individuals (dataset 1, 12,756 SNPs). Prior filtering for linkage
450 disequilibrium is not necessary for analyses using the RADpainter tool as it efficiently
451 estimates the effective number of loci in mapped data files during the analysis [33]. The
452 fineSTRUCTURE [80] clustering algorithm was run with a burn-in of 100,000 iterations
453 followed by 100,000 sampled iterations and the tree building algorithm was run with a burn-
454 in of 10,000 iterations. We then performed a principal coordinate analysis (PCoA) on the
455 linkage filtered dataset (dataset 2, 9668 SNPs). The PCoA was performed using Elucidean

456 distances with the package *adegenet* in R. VCF files were converted to *genpop* format for
457 input to *adegenet* using *PGDSpider*, version 2.1.1.5 [81].

458 *Phenotypic divergence*

459 Genomic analyses grouped all fish into two broad genomic clusters, cluster 1 and cluster 2.
460 Although marine fish were grouped with the freshwater fish in cluster 1, they were treated as
461 a third, separate, group in all further analyses. Additionally, they were excluded from most
462 subsequent genetic analyses because their presence in freshwater/coastal areas is transient
463 (they migrate to shallow coastal areas only in the spring to breed) and they represent the
464 likely ancestral state of all freshwater populations [82]. To determine the phenotype of these
465 three groups (marine, cluster 1 and cluster 2), we calculated the mean of each phenotypic
466 variable (weight, number of gill rakers, mean raker length, armour PC1 and shape PC1) for
467 each group. To test whether the means of each phenotypic variable in each of the three
468 groups were significantly different from one another, linear mixed models were performed
469 using the *nlme* package [83], with lake included as a random effect and group (marine,
470 cluster 1, cluster 2) as a fixed effect. For models showing a significant effect of group, post-
471 hoc pairwise comparisons were performed using estimated marginal means, implemented
472 using the *emmeans* package [84] in R. P-values for post-hoc comparisons were adjusted for
473 multiple testing using the False Discovery Rate (FDR) method [85].

474 *Phylogenetic analyses*

475 Prior to phylogenetic analysis, we filtered our master dataset (dataset 1) to remove all known
476 QTL in stickleback. Data for QTL were downloaded from Peichel and Marques [59],
477 converted to BED format, and removed from the VCF file using *VCFtools*. This reduced the
478 number of SNPs from 12,756 to 8351. We then ensured approximate linkage equilibrium of

479 remaining markers by removing all SNPs with an R^2 value >0.2 using Plink version 1.9. This
480 left 6215 SNPs, dataset 3.

481 To construct a phylogeny of all sequenced individuals, we used a bootstrapped maximum-
482 likelihood based approach, implemented in RAxML, version 8.2.12 [86]. The VCF file was
483 converted to phylip format for input to RAxML using python version 3.8.2. RAxML was run
484 with a GTR-GAMMA model of substitution rate heterogeneity, automatic bootstrap replicate
485 halting using the autoMRE function and with the default settings for all other parameters.

486 To assess the robustness of the maximum likelihood phylogeny, we also performed topology
487 weighting using TWISST [87]. Topology weighting was carried out on four freshwater
488 populations, with a single marine sequence (from LICA) as the outgroup. The freshwater
489 populations were selected to contain two pairs of geographically proximal populations, with
490 one pair from either side of the Georgia strait (NORT and BEAV, and DOUG and KLEN),
491 and with one population from each pair falling in cluster 1, the other in cluster 2. Dataset 2
492 (9668 SNPs) was filtered to contain all individuals from each of the four freshwater
493 populations and a single individual from LICA (TWISST only accepts a single sequence as
494 an outgroup), using VCFtools (dataset 4, 9668 SNPs). The VCF file was converted to .geno
495 format and maximum likelihood trees were estimated in phym1 [88] in sliding windows of
496 50bp using Python 2.7.15 and the scripts available with TWISST. Topology weightings were
497 then computed using Python 3.8.2 and topologies were visualised in R.

498 *Phylogenetic signal in phenotypic traits*

499 To estimate phylogenetic signal, the phylogeny constructed in RAxML was imported into R
500 using the ape package [89], individual nodes were collapsed to leave a single node per
501 population, with the exception of the two marine populations, which were both collapsed into
502 one node using the phytools [90] and phangorn [91] R packages. Phenotypic trait data

503 (weight, number of gill rakers, mean raker length, armour PC1 and shape PC1) were added to
504 the tree tips, and phylogenetic signal and associated p -values for each trait were estimated
505 using the package phylosignal [92]. We used Pagel's λ [93] to estimate phylogenetic signal as
506 this statistic performs well compared to others available and has a low type 1 error rate [39,
507 94, 95]. P -values are calculated using likelihood ratio tests that compare the observed λ
508 statistic with a phylogenetically independent trait distribution. We also simulated data for five
509 additional traits to be distributed randomly with regard to phylogeny. Simulated traits were
510 tested alongside the real phenotypic variables for comparison.

511 As we aim to detect whether benthic and limnetic characteristics have evolved a single time
512 or repeatedly across the radiation, we also used the R package SURFACE [32] to estimate the
513 most likely number of different selection regimes (k) and instances of convergent evolution
514 (c) by identify the best fitting model of trait evolution for our phylogeny and associated
515 phenotypic traits. SURFACE begins by fitting a single peak Ornstein-Uhlenbeck (OU) model
516 (which allows for a single adaptive optimum and variation in the parameter α , which
517 describes the strength of selection towards that optimum) by maximum likelihood. It then
518 sequentially adds adaptive peaks to the model in a step-wise process and accepts each more
519 complex model until AIC values are no longer improved. SURFACE then attempts to
520 collapse regimes with the same optima in a process of step-wise backwards selection
521 whereby if multiple optima are the same, the AIC of the model is improved by reducing the
522 number of model parameters. We ran SURFACE using the same collapsed phylogeny and
523 associated trait data that was generated to estimate phylogenetic signal. The tree was
524 converted to ouchtree format and the best fitting model of trait evolution was estimated under
525 an AIC threshold of 0 (any improvement in AIC should be accepted) using the SURFACE R
526 package.

527 *Relationship between phenotype and environment*

528 To investigate associations between environmental characteristics and divergence in
529 phenotypic traits, we used a phylogenetic generalised least squares (PGLS) approach so that
530 phylogenetic signal could be accounted for in the models, using the ape [89], nlme [96] and
531 geiger [97] packages in R. Marine fish were excluded from all phenotype – environment
532 modelling because our main aim was to detect effects in relation to the freshwater benthic
533 and limnetic phenotypes in BC and, although marine fish have a limnetic phenotype, we
534 found them to differ phenotypically from the freshwater limnetic fish in BC. Separate models
535 were run for each phenotypic trait (weight, number of gill rakers, mean gill raker length,
536 armour PC1 and shape PC1). Models were fitted by maximum likelihood and we began with
537 all environmental variables in each model (mean lake depth [m], lake area [km²],
538 presence/absence of prickly sculpin, pH and calcium concentration [mg/L]). Terms were then
539 removed sequentially, with the least significant terms removed first, until the reduced model
540 was no longer a significant improvement on the fuller model under the $p < 0.05$ threshold.
541 Model comparison was conducted using Wald tests. Phylogenetic effects for each phenotypic
542 trait were accounted for in each model following the principles set out in Mazel et al. [98].
543 We first transformed the phylogeny for each phenotypic trait under a lambda model with
544 lambda specified as the lambda estimate for that phenotypic trait in the phylogenetic signal
545 analyses. The phylogenetic variance-co-variance matrices were computed from the
546 transformed trees using the ape package and converted to correlation matrices, which were
547 used to specify phylogenetic correlation of errors in the models.

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- 836

837 **Supplementary Tables**838 **Table S1**

839 **Description of sample sites.** Sample sizes are shown for phenotypic (n pheno) and RAD (n RAD) analyses. Sal refers to salinity classifications,
 840 Fr: freshwater (absolute conductivity <500 $\mu\text{S}/\text{cm}$), Br: brackish (absolute conductivity 20,000-35,000 $\mu\text{S}/\text{cm}$). Sampling locations are given by
 841 latitude followed by longitude.

Lake ID	Lake name	n pheno	n RAD	Genomic cluster	Mean depth (m)	Area (km ²)	Sculpin	Sal	Location
AMBR	Ambrose lake	29	17	1	13.3	0.2980	Yes	Fr	49°44'3"N; 124°1'16"W
BEAV	Beaver lake	30	19	1	3.5	0.1932	Yes	Fr	48°48'42"N; 124° 4'51"W
BRAN	Brannen lake	30	19	2	11.6	1.0866	Yes	Fr	49°12'55"N; 124°3'18"W
BULL	Bullocks lake	30	17	2	4.0	0.0940	No	Fr	48°52'25"N; 123°30'33"W
CRAN	Cranby lake	30	20	2	3.2	0.4460	No	Fr	49°41'36"N; 124°30'42"W
DOUG	Dougan lake	30	16	2	8.5	0.1000	No	Fr	48°42'53"N; 123°36'48"W
GARD	Garden bay	10	9	1	10.0	0.6236	Yes	Fr	49°39'1"N; 124°1'23"W
HOGG	Hoggan lake	30	20	2	3.0	0.1970	No	Fr	49°9'7"N; 123°49'43"W
HOTE	Hotel	35	17	1	5.9	0.2522	No	Fr	49°38'26"N; 124°3'3"W
KENN	Kennedy lake	30	19	1	38.0	65.0000	Yes	Fr	49°3'59"N; 125°28'2"W
KIRK	Kirk lake	30	18	2	8.3	0.0830	No	Fr	49°44'22"N; 124°34'57"W
KLEN	Klein lake	30	17	2	12.0	0.1350	No	Fr	49°43'53"N; 123°58'23"W
LICA	Little Campbell river	34	10	1	3.0	74.400	Yes	Br	49°0'51"N; 122°45'32"W
LILY	Lily lake	30	19	1	2.2	0.1214	Yes	Fr	49°36'44"N; 124° 1'17"W
NORT	North lake	14	14	1	10.1	0.1279	Yes	Fr	49°44'60"N; 123°58'27"W
OYST	Oyster lagoon	30	10	1	1.3	0.0200	Yes	Br	49°36'49"N; 124°1'47"W
PAXT	Paxton lake	63	9	1 / 2	6.2	0.1700	No	Fr	49°42'22"N; 124°31'24"W
PRIE	Priest lake	57	9	1 / 2	5.4	0.4434	No	Fr	49°44'50"N; 124°33'52"W
SPRO	Sproat lake	30	19	1	65.5	37.7500	Yes	Fr	49°16'0"N; 125°2'13"W
STOW	Stowell lake	30	17	2	4.6	0.0564	No	Fr	48°46'54"N; 123°26'37"W
TROU	Trout lake	30	18	2	5.8	0.0756	No	Fr	49°30'26"N; 123°52'37"W

842

843 **Table S2**

844 **SNPs in linkage cluster 10 that lie within genes.** Genome location of SNPs falling within
 845 coding regions and the associated gene and gene ID for each SNP.

Chromosome	SNP	Gene name	Ensembl gene ID
groupI	138067		ENSGACG00000004458
groupI	1913469		ENSGACG00000005744
groupI	14210802		ENSGACG00000011908
groupI	22123269		ENSGACG00000014464
groupII	20248077	NELL1	ENSGACG00000017098
groupIV	4186906	pde6a	ENSGACG00000016811
groupV	291741	kat6b	ENSGACG00000002173
groupV	1957242	glud1b	ENSGACG00000002813
groupV	2216160	cyth1b	ENSGACG00000002991
groupVI	2259976	cpeb3	ENSGACG00000002944
groupVI	6422836	fbxolla	ENSGACG00000005432
groupVI	13171573	sdccag8	ENSGACG00000010549
groupVII	20193529	dclk1a	ENSGACG00000020560
groupX	8795111	fam126a	ENSGACG00000006077
groupXIII	9178337	ube3b	ENSGACG00000009166
groupXIII	15603103	arvcfb	ENSGACG00000013003
groupXIII	17477236	krt1-c5	ENSGACG00000013907
groupXIV	4013936	dnajc21	ENSGACG00000016480
groupXV	9968890	ppp2r3a	ENSGACG00000010930
groupXV	14295031	odc1	ENSGACG00000012974
groupXVI	119418		ENSGACG00000001436
groupXVI	210791	pde9a	ENSGACG00000001450
groupXVII	840001	ccdc51	ENSGACG00000003571
groupXVII	6327443	trim62.1	ENSGACG00000007462
groupXXI	2692493	nrros	ENSGACG00000002094
scaffold_114	90942	itpr3	ENSGACG00000000253
scaffold_114	90942		ENSGACG00000000233
scaffold_27	3558934	wdr82	ENSGACG00000001254

846

847 **Table S3**

848 **Linear mixed model results for phenotypic differences between groups.** Table shows the
849 results of linear mixed models testing for differences between three groups (marine, genomic
850 cluster 1 and genomic cluster 2) in five phenotypic traits, with population as a random effect.

Trait	<i>F</i>	<i>df</i>	<i>p</i>-value
Weight	4.4219	2	0.0257
N gill rakers	5.4898	2	0.0126
Raker length	12.4074	2	0.0003
Armour PC1	8.1948	2	0.0025
Shape PC1	23.0643	2	<0.0001

851 *F* describes the likelihood ratio test statistic, *df* describes degrees of freedom. *P*-values < 0.05
852 are highlighted in bold.

853 **Table S4**

854 **Post-hoc Estimated marginal means test results for pairwise phenotypic differences**
 855 **between groups.** Table shows the results of post-hoc estimated marginal means tests to
 856 determine pairwise differences between groups (marine, genomic cluster 1 and genomic
 857 cluster 2) when phenotypes differed significantly between groups in linear mixed models.

Trait	Groups compared	Estimate	SE	<i>df</i>	Adjusted <i>p</i> -value
Weight	limnetic-like – benthic-like	0.2453	0.2400	20	0.3188
	marine – benthic-like	-1.0119	0.4193	20	0.0383
	marine – limnetic-like	-1.2572	0.4232	20	0.0227
N gill rakers	limnetic-like – benthic-like	-2.0254	0.6387	20	0.0144
	marine – benthic-like	-1.9978	1.1211	20	0.1349
	marine – limnetic-like	0.0276	1.1304	20	0.9808
Raker length	limnetic-like – benthic-like	-0.1799	0.0639	20	0.0107
	marine – benthic-like	-0.5286	0.1121	20	0.0004
	marine – limnetic-like	-0.3487	0.1130	20	0.0088
Armour PC1	limnetic-like – benthic-like	-2.2551	0.6914	20	0.0069
	marine – benthic-like	-3.8791	1.2155	20	0.0069
	marine – limnetic-like	-1.6240	1.2250	20	0.1999
Shape PC1	limnetic-like – benthic-like	-2.4654	0.4697	20	<0.0001
	marine – benthic-like	-4.5652	0.8202	20	<0.0001
	marine – limnetic-like	-2.0998	0.8280	20	0.0197

858 Estimates describe the mean difference between groups, SE describes the standard error of
 859 the estimates and *df* describe degrees of freedom. *P*-values are adjusted for multiple testing
 860 using the FDR method. *P*-values < 0.05 are highlighted in bold.

861 **Table S5**

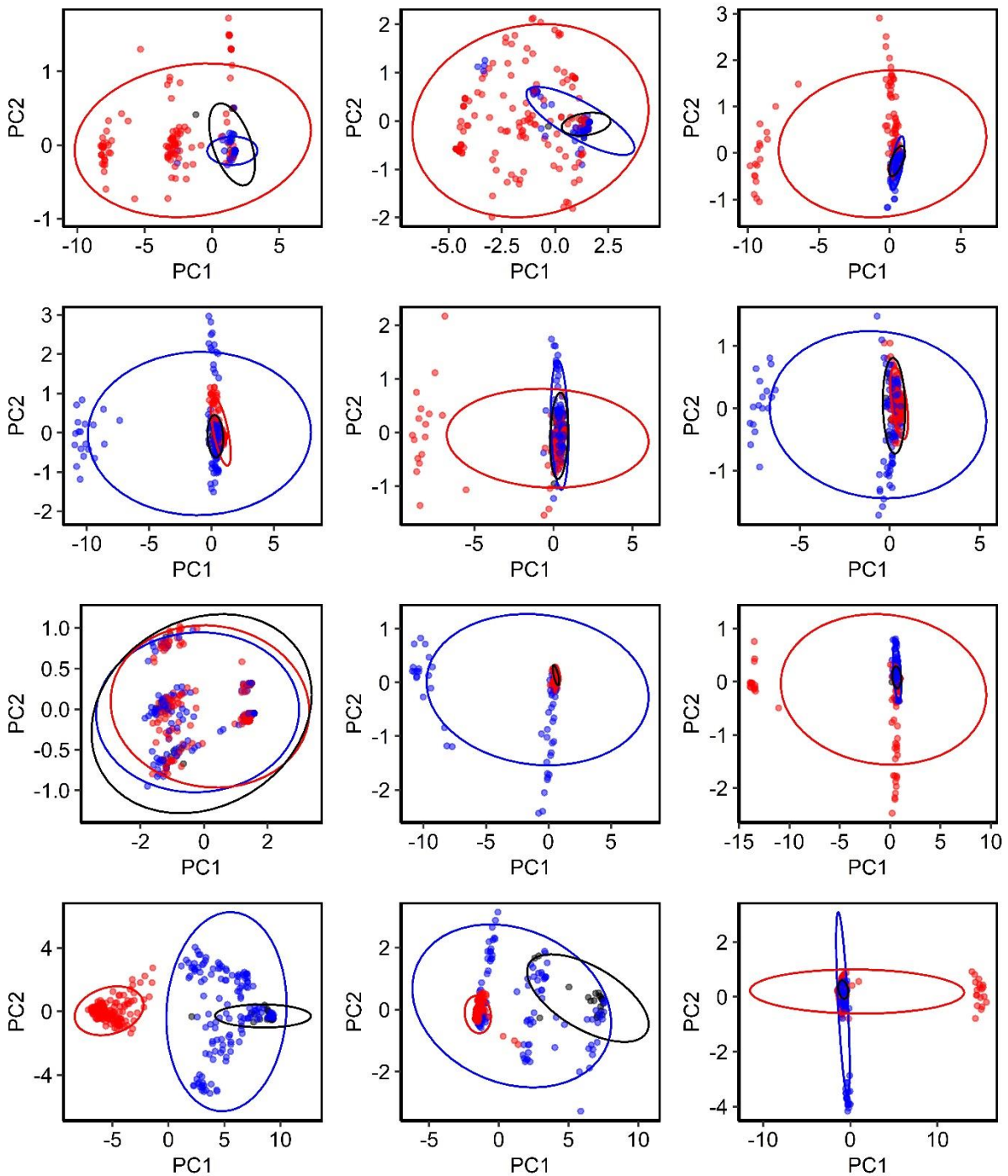
862 **Effects of the environment on the distribution of benthic-limnetic phenotypic traits.**

863 Table shows the results of phylogenetic generalised least squares analyses (PGLS) on each
 864 phenotypic trait. *P*-values < 0.05 are highlighted in bold. *P*-values were adjusted for multiple
 865 testing using the FDR method.

Phenotypic trait	Environmental variable	<i>df</i>	Wald statistic	<i>p</i> -value	Adjusted <i>p</i> -value
Armour PC1	Lake area	1	1.5877	0.2077	0.5077
	Ca concentration	1	3.2274	0.0724	0.2172
	Mean lake depth	1	0.0883	0.7663	0.8969
	pH	1	1.6609	0.1975	0.3950
	Sculpin presence	1	17.6411	0.0000	0.0002
N rakers	Lake area	1	0.0461	0.8300	0.8300
	Ca concentration	1	0.3525	0.5527	0.6088
	Mean lake depth	1	0.0327	0.8564	0.8969
	pH	1	0.1860	0.6663	0.7882
	Sculpin presence	1	0.7284	0.3934	0.4721
Raker length	Lake area	1	6.8857	0.0087	0.0521
	Ca concentration	1	1.4304	0.2317	0.4634
	Mean lake depth	1	0.0168	0.8969	0.8969
	pH	1	0.5091	0.4755	0.7133
	Sculpin presence	1	2.6932	0.1008	0.1512
Shape PC1	Lake area	1	0.0641	0.8001	0.8300
	Ca concentration	1	0.2620	0.6088	0.6088
	Mean lake depth	1	0.1346	0.7137	0.8969
	pH	1	1.7163	0.1902	0.3950
	Sculpin presence	1	3.8490	0.0498	0.0996
Weight	Lake area	1	0.1549	0.6939	0.8300
	Ca concentration	1	0.5770	0.4475	0.6088
	Mean lake depth	1	16.8205	0.0002	0.0013
	pH	1	7.9713	0.0048	0.0285
	Sculpin presence	1	0.0013	0.9714	0.9714

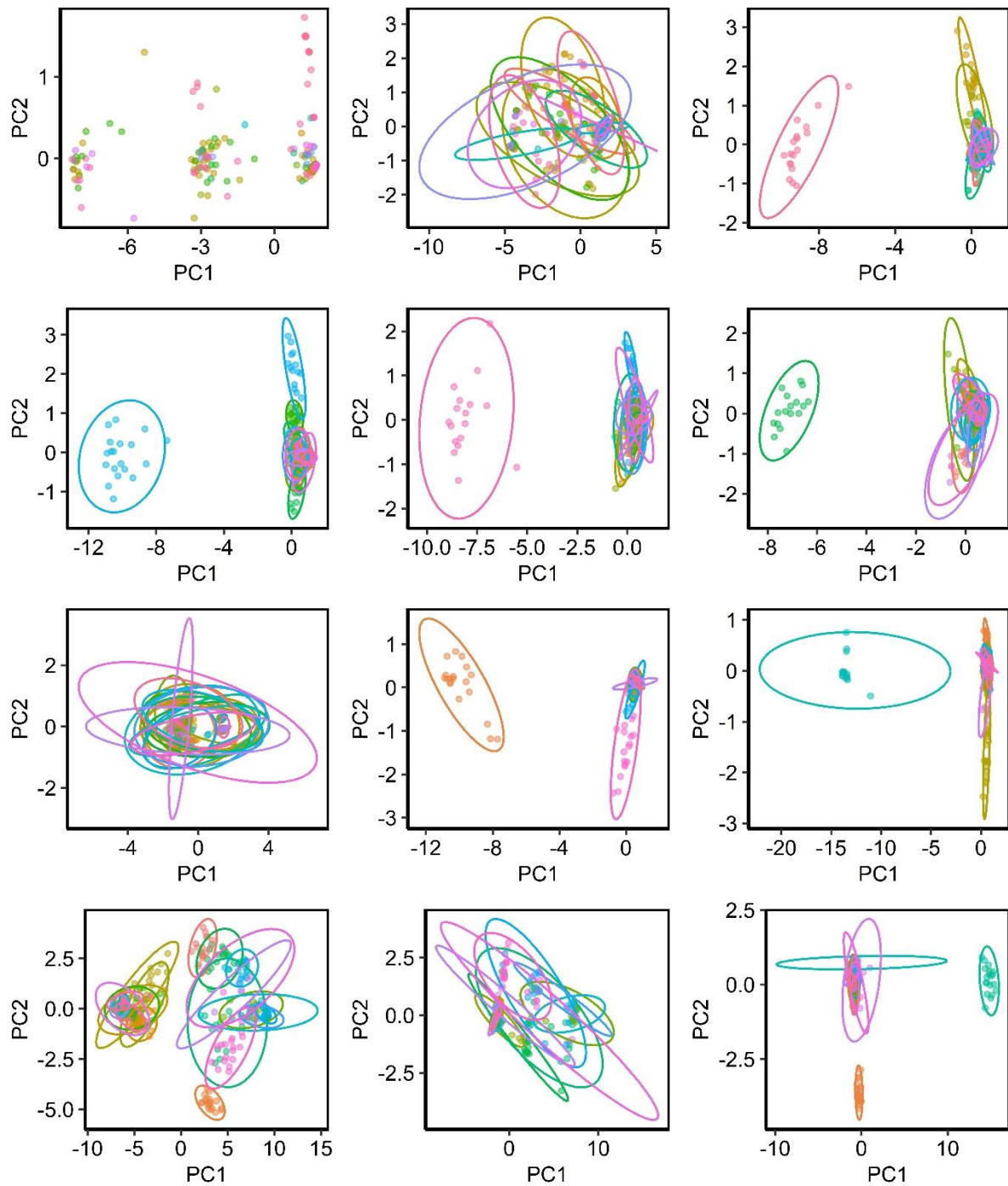
866

867 **Supplementary Figures**



868

869 **Figure S1.** Principal component analyses (PCA) of each LD cluster in the LD network
870 identified by LDna. Black circles represent marine individuals, blue circles, freshwater
871 cluster 1 and red circles freshwater cluster 2. 95% confidence ellipses are shown for each
872 PCA in the corresponding colours.



873

874 **Figure S2.** Principal component analyses (PCA) of each LD cluster in the LD network
875 identified by LDna. Coloured circles and ellipses represent each of the 23 sampled
876 populations.