

## Genetic models reveal historical patterns of sea lamprey population fluctuations within Lake Champlain

Cassidy C D'Aloia, Christina B Azodi, Sallie P Sheldon, Stephen C Trombulak, Willam R Ardren

The origin of sea lamprey (*Petromyzon marinus*) in Lake Champlain has been heavily debated over the past decade. Given the lack of historical documentation, two competing hypotheses have emerged in the literature. First, it has been argued that the relatively recent population size increase and concomitant rise in wounding rates on prey populations are indicative of an invasive population that entered the lake through the Champlain Canal. Second, recent genetic evidence suggests a post-glacial colonization at the end of the Pleistocene, approximately 11,000 years ago. One limitation to resolving the origin of sea lamprey in Lake Champlain is a lack of historical and current measures of population size. In this study, the issue of population size was explicitly addressed using nuclear (nDNA) and mitochondrial DNA (mtDNA) markers to estimate historical demography with genetic models. Haplotype network analysis, mismatch analysis, and summary statistics based on mtDNA noncoding sequences for NCI (479 bp) and NCII (173 bp) all indicate a recent population expansion. Coalescent models based on mtDNA and nDNA identified two potential demographic events: a population decline followed by a very recent population expansion. The decline in effective population size may correlate with land-use and fishing pressure changes post-European settlement, while the recent expansion may be associated with the implementation of the salmonid stocking program in the 1970s. These results are most consistent with the hypothesis that sea lamprey are native to Lake Champlain; however, the credibility intervals around parameter estimates demonstrate that there is uncertainty regarding the magnitude and timing of past demographic events.

1 **TITLE:**

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31 **ABSTRACT**

32 The origin of sea lamprey (*Petromyzon marinus*) in Lake Champlain has been heavily debated  
33 over the past decade. Given the lack of historical documentation, two competing hypotheses  
34 have emerged in the literature. First, it has been argued that the relatively recent population size  
35 increase and concomitant rise in wounding rates on prey populations are indicative of an invasive  
36 population that entered the lake through the Champlain Canal. Second, recent genetic evidence  
37 suggests a post-glacial colonization at the end of the Pleistocene, approximately 11,000 years  
38 ago. One limitation to resolving the origin of sea lamprey in Lake Champlain is a lack of  
39 historical and current measures of population size. In this study, the issue of population size was  
40 explicitly addressed using nuclear (nDNA) and mitochondrial DNA (mtDNA) markers to  
41 estimate historical demography with genetic models. Haplotype network analysis, mismatch  
42 analysis, and summary statistics based on mtDNA noncoding sequences for NCI (479 bp) and  
43 NCII (173 bp) all indicate a recent population expansion. Coalescent models based on mtDNA  
44 and nDNA identified two potential demographic events: a population decline followed by a very  
45 recent population expansion. The decline in effective population size may correlate with land-use  
46 and fishing pressure changes post-European settlement, while the recent expansion may be  
47 associated with the implementation of the salmonid stocking program in the 1970s. These results  
48 are most consistent with the hypothesis that sea lamprey are native to Lake Champlain; however,  
49 the credibility intervals around parameter estimates demonstrate that there is uncertainty  
50 regarding the magnitude and timing of past demographic events.

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53

54 **INTRODUCTION**

55 The origin of the landlocked population of sea lamprey (*Petromyzon marinus*) in Lake  
56 Champlain has been the subject of an ongoing debate in recent years (Bryan et al., 2005;  
57 Waldman, Grunwald & Wirgin, 2006; Waldman et al., 2009; Eshenroder, 2009, 2014). The sea  
58 lamprey is an anadromous fish that has a parasitic juvenile phase during which it feeds on the  
59 bodily fluids of a variety of prey fishes, including large salmonids such as lake trout (*Salvelinus*  
60 *namaycush*), Atlantic salmon (*Salmo salar*), and lake whitefish (*Coregonus clupeaformis*), as  
61 well as lake sturgeon (*Acipenser fulvescens*). Recent research has suggested that native coastal  
62 populations of sea lamprey may have positive environmental impacts on freshwater streams. For  
63 example, post-spawning sea lamprey carcasses may be important sources of marine-derived  
64 nutrients and materials in oligotrophic streams (Guyette et al., 2014). However, the overall effect  
65 of landlocked sea lamprey populations in the Great Lakes and Lake Champlain has been  
66 detrimental. Lamprey-induced collapses of native fish populations have been well documented in  
67 the region since the 1970s (Smith, 1971; Smith & Tibbles, 1980).

68         Consequently, control efforts were developed in the Great Lakes to suppress sea lamprey  
69 populations and facilitate restoration of native species. Control methods are varied and include  
70 widespread biocide use, physical migration barriers, and spawning-phase traps in tributaries.  
71 Although these methods have been criticized for the potential negative effects that non-target  
72 species may experience (McLaughlin, Marsden & Hayes, 2003), the effort has been largely  
73 successful in terms of lamprey control. Using the methods developed in the Great Lakes, an  
74 experimental lamprey control program was implemented in Lake Champlain in 1990, followed  
75 by a long-term program beginning in 2001. Although the control methods in Lake Champlain are  
76 similar to those in the Great Lakes, a major difference is the consensus regarding the fish's status

77 as native versus invasive; in the Great Lakes, sea lamprey are known to be invasive (with the  
78 exception of Lake Ontario), while much debate surrounds the population in Lake Champlain.

79         To date, two alternative historical scenarios have dominated the discussion regarding the  
80 origin of sea lamprey in Lake Champlain. First, it has been argued that individuals from the  
81 Atlantic coast population invaded Lake Champlain via the canal system sometime between the  
82 1840s and the 1920s (Eshenroder, 2009, 2014). This hypothesis is based on the fact that the first  
83 documentation of the species in the lake was in 1929 (Eshenroder, 2014), 13 years after the  
84 construction of the Champlain Barge Canal (the third version of the Champlain Canal), which  
85 directly connected Lake Champlain to the Hudson River. Second, recent genetic analyses based  
86 on both nuclear (nDNA) and mitochondrial DNA (mtDNA) markers suggest that sea lamprey are  
87 native to Lake Champlain (Bryan et al., 2005; Waldman, Grunwald & Wirgin, 2006; Waldman et  
88 al., 2009). Bryan et al. (2005) tested alternative coalescent-based colonization models using  
89 microsatellite markers and found evidence of long-term vicariance in the Lake Champlain  
90 population, concluding that the most probable route of entry was via the St. Lawrence River  
91 upon the initial formation of Lake Champlain about 12,500 years ago. Waldman et al. (2006)  
92 compared haplotype frequencies of the Lake Champlain population to Atlantic Coast and Great  
93 Lakes populations using mitochondrial non-coding DNA and concluded that the data were most  
94 consistent with a post-glacial colonization during a period when modern-day Lake Champlain  
95 was an arm of the Atlantic Ocean called the Champlain Sea. However, neither hypothesis has  
96 been widely accepted throughout the scientific and management communities. The “nonnative”  
97 hypothesis has been criticized because it is based on the absence of species documentation data  
98 during the 1800s — a time when systematic biological censuses were not conducted in the lake  
99 — and because it lacks due consideration of the post-glacial geological history of the region.

100 Likewise, the “native” hypothesis has recently been called into question after an extensive  
101 review of historical documentation suggested that the time of origin used in genetic models  
102 (1841) was based on an erroneous species identification and, subsequently, may have biased the  
103 results (Eshenroder, 2014).

104         As a result of this ongoing debate, prior research has focused exclusively on the timing of  
105 the origin of *P. marinus* in Lake Champlain, while little is known about the rest of the  
106 population’s history. The two previous genetic studies of *P. marinus* in the lake were regional  
107 studies that compared the Lake Champlain sea lamprey population to other populations from the  
108 Great Lakes and the Atlantic Ocean (Bryan et al., 2005; Waldman, Grunwald & Wirgin, 2006).  
109 While both studies found evidence for long-term vicariance from anadromous and other  
110 freshwater populations, it has proven challenging to reconcile the results with the complete lack  
111 of historical documentation of sea lamprey in the Lake Champlain (Bryan et al., 2005). Given  
112 that these regional studies have already shown the Lake Champlain population to be  
113 differentiated from all other populations, additional analyses can further probe the genetic data  
114 by conducting rigorous intra-population analyses to model historical population dynamics *within*  
115 Lake Champlain. An objective assessment of the timing and magnitude of fluctuations in  
116 population size over time may therefore provide a more complete understanding of the history of  
117 *P. marinus* in Lake Champlain, while simultaneously shedding new light on the contentious  
118 topic of whether or not the species is invasive.

119         In cases such as this one, where historical census data are unavailable, genetic markers  
120 can be powerful tools for inferring demographic fluctuations. These inferences are possible  
121 because the census population size ( $N_c$ ) is generally proportional to the effective population size  
122 ( $N_e$ ). Effective population size is the size of an idealized population (i.e., with binomial variance

123 in reproductive success, an equal sex ratio, and discrete generations), that is subject to the same  
124 level of genetic drift and inbreeding as the census population (Wright, 1938). Drawing on the  
125 relationship between effective and census population size, and the fact that rapid demographic  
126 fluctuations can be detected with genetic markers, we can investigate how effective population  
127 size has changed over time and, in turn, infer proportional changes in the overall population  
128 (Waples, 1989; Frankham, 1995).

129         Traditionally, genetic approaches to estimating historical demography have used  
130 summary statistics to test whether extant population-level data deviate from theoretical  
131 expectations under alternative models of population stasis, contractions, and expansions (Cornuet  
132 & Luikart, 1996; Harpending et al., 1998; Schneider & Excoffier, 1999; Garza & Williamson,  
133 2001). For example, mismatch analysis uses sequence data to compare the distribution of  
134 observed pairwise differences between all haplotypes in a population to the distribution expected  
135 under a specified population change. Expansion, contractions, and equilibrium each generate a  
136 particular pattern of the distribution of pairwise differences among sequences. Contractions or  
137 equilibrium lead to multimodal, ragged distributions while expansions result in a smooth  
138 unimodal Poisson distribution of pairwise differences (Harpending et al., 1998; Schneider &  
139 Excoffier, 1999). These moment-based metrics are widely used because they are easy to obtain  
140 with sequence and/or allele frequency data. However, they provide crude approximations of  
141 population changes, and their precision is linked to the timing and magnitude of the demographic  
142 change in question.

143         Arguably, a more powerful approach to inferring past demographic change is coalescent  
144 modeling (Storz & Beaumont, 2002; Beaumont & Rannala, 2004). Coalescent theory seeks to  
145 describe the ancestral relationship of a particular gene or set of genes by recognizing that the

146 probability of two lineages coalescing during a particular generation is inversely proportional to  
147 effective population size at that time (Beaumont & Rannala, 2004; Kuhner, 2008;). Thus, these  
148 models trace separate genetic lineages back to their most recent common ancestor (Kuhner,  
149 2008) and connect these genealogies to changes in effective population size (Storz & Beaumont,  
150 2002). Models such as BEAST and Msva adopt Bayesian Markov chain Monte Carlo (MCMC)  
151 methods to explore parameter space and sample the posterior distributions of the demographic  
152 parameters of interest (Beaumont, 1999; Drummond & Rambaut, 2007). A key advantage to  
153 these methods is the ability to include time as one of the estimated parameters, as opposed to  
154 using fixed time points as assumptions in the model. Directly estimating time is particularly  
155 important for models of historical sea lamprey demography, as previous genetic studies that  
156 focused on the population's origin have been criticized for using fixed, potentially-incorrect  
157 dates (Eshenroder, 2014).

158         The sea lamprey is a tractable study species for coalescent modeling of effective  
159 population size fluctuations because several genetic resources are available. The entire  
160 mitochondrial genome is sequenced (Lee & Kocher, 1995), enabling researchers to sequence  
161 hypervariable regions of mitochondrial DNA (mtDNA). For decades, mtDNA has been widely  
162 used to infer demographic processes because of its maternal inheritance, small effective  
163 population size, and relatively fast rate of evolution (Avice, 1994; White et al., 2008). These  
164 characteristics are particularly useful for demographic studies because uniparental inheritance  
165 can be modeled without the complications of recombination, and signatures of relatively recent  
166 demographic events are more readily detectable when effective population size is small and  
167 mutation rate is elevated. Also, *P. marinus* has two non-coding regions in the mitochondrial  
168 genome: Non-coding region one (NCI) is 491-bp long and non-coding region two (NCII) is 199-

169 bp long (Lee & Kocher, 1995). Finally, a suite of microsatellite markers has already been  
170 developed for this species (Bryan et al., 2005); thus, markers from the mitochondrial and nuclear  
171 genomes can be used concurrently to study the population's history. The concurrent use of two  
172 genomic marker types enables intra-population replication for inferring demographic history  
173 (Eytan & Hellberg, 2010).

174         The purpose of this study was to use multiple analytical approaches and two sets of  
175 genetic markers to investigate historical population fluctuations in Lake Champlain sea lamprey.  
176 First, moment-based methods — including a mismatch distribution and Fu's  $F_S$  statistic — were  
177 used to generate coarse estimates of historical population expansions and/or contractions based  
178 on mtDNA sequence data (NCI and NCII). Second, two coalescent MCMC models were used to  
179 explicitly estimate changes in effective population size over time. Mitochondrial sequence data  
180 (NCII) were used to generate a Bayesian Skyline Plot (BSP) in the program BEAST to model  
181 effective population size history while taking into account coalescent and phylogenetic  
182 uncertainty. Previously-published allele frequency data for eight nuclear microsatellite loci  
183 (Bryan et al., 2005) were also used in the program Msvar to estimate four demographic  
184 parameters: historical effective population size, current effective population size, mutation rate,  
185 and time. In total, we used 10 loci ( $n = 2$  non-coding mtdna;  $n=8$  nuclear microsatellites), and  
186 employed both moment-based methods as well as two classes of coalescent models to explore  
187 historical demography. Taken together, these data can provide insight into signatures of  
188 demographic events within two separate genomes.

189

## 190 **METHODS**

191 *Sample collection and mitochondrial DNA sequencing*

192 To estimate historical population fluctuations within Lake Champlain, fin tissue samples from  
193 spawning-phase sea lamprey were obtained from the US Fish and Wildlife Service (USFWS) in  
194 May-June 2009. All tissue collection was conducted as part of routine USFWS sampling; the  
195 authors of this study received tissue, but did not handle any vertebrate specimens. First, to  
196 estimate fluctuations based on mtDNA sequence data, samples were collected at three Lake  
197 Champlain tributaries: Great Chazy River (n = 33), Malletts Creek (n = 33), and Beaver Brook (n  
198 = 28), representing the northern, central, and southern regions of Lake Champlain, respectively.  
199 A broad geographic sampling regime was used in order to test for population structure. All tissue  
200 samples were stored in 95% non-denatured ethanol and genomic DNA was extracted using  
201 DNeasy Blood and Tissue Kits (Qiagen).

202 The two non-coding regions of the mtDNA genome, NCI and NCII, were amplified in all  
203 Lake Champlain samples using PCR with the lamprey-specific primers CR1 (Waldman et al.,  
204 2004) and LampR (5'- AATAGACGGTTGGTGGGACA - 3'). PCR reactions were performed  
205 in 25 µl volumes with the following reagents: 0.2 µl Qiagen *taq* polymerase (5 units/ µl), 2.5 µl  
206 10X PCR buffer (with 1.5 mM MgCl<sub>2</sub>), 10 µM each primer, 10 µM dNTP, and 50-100 ng  
207 template DNA. Thermal cycler settings were set at an initial denaturation at 95°C for 5 minutes  
208 followed by 40 cycles of 95° C for 45 seconds, 56.5° C for 45 seconds, and 72° C for 1 minute;  
209 and a final extension at 72° C for 10 minutes. PCR products were cleaned using the Wizard SV  
210 Gel and PCR Clean-Up Kit (Promega) and eluted in 25 µl H<sub>2</sub>O.

211 To sequence the Lake Champlain samples at both non-coding regions, forward extension  
212 reactions were run for each individual using the forward primers CR1 (for NCI) and CASSFN  
213 (for NCII) (5'- GACCCCTAAGTTCATTGC - 3'). All primers designed specifically for this  
214 study, including LampR, CR1, and CASSFN, were designed using Primer3 (Untergrasser et al.,

215 2012). Extension reactions were prepared using  $\frac{1}{4}$  reactions from the BigDye Terminator v3.1  
216 Cycle Sequencing Kit (Applied Biosystems). For each 20  $\mu$ l reaction, the following reagents  
217 were combined: 2  $\mu$ l Ready Reaction Mix, 3  $\mu$ l 5x sequencing buffer, 3.2  $\mu$ l primer (1  $\mu$ M), 1  $\mu$ l  
218 DNA template (20-30 ng), and 10.8  $\mu$ l water. Thermal cycler settings were set at an initial  
219 denaturation at 96° C for 1 minute followed by 30 cycles of 96° C for 20 seconds, 50° C for 20  
220 seconds, and 60° C for 4 minutes. Reactions were cleaned with 2  $\mu$ l of 2.2% sodium dodecyl  
221 sulfate and returned to the thermal cycler at 98° C for 5 minutes followed by 25° C for 10  
222 minutes. Reactions were then purified in Sephadex size-exclusion columns (GE Healthcare Life  
223 Sciences) and 12  $\mu$ l of purified products were loaded into an Applied Biosystems 3130 Genetic  
224 Analyzer for sequencing.

225         Sequences from individuals that successfully amplified at both non-coding regions  
226 (n=54) were concatenated into a single sequence denoted by NC<sub>total</sub>. Concatenated sequences  
227 were then trimmed to a 652-bp region to exclude repetitive portions of reads prone to slippage  
228 (NCI: excluded 14 bp before position 15382; NCII: excluded 28 bp after position 16173).  
229 Regions prone to slippage were identified by superimposed sequences on the chromatograms.  
230 All trimmed sequences were aligned with the MUSCLE algorithm (Edgar, 2004). Finally, a  
231 haplotype network was generated for all NC<sub>total</sub> haplotypes to visualize their relationship using  
232 TCS v. 1.18 (Clement, Posada & Crandall, 2000)

233         Second, to estimate population fluctuations based on nuclear microsatellite data, we used  
234 previously-published allele frequency data (Bryan et al., 2005). Bryan et al. (2005) genotyped  
235 individuals from Great Chazy River (n=40) and Lewis Creek (n=40) at 8 microsatellite loci.  
236 There was no evidence for linkage disequilibrium or significant deviations from Hardy-  
237 Weinberg equilibrium (see Bryan et al., 2005 for more details).

238

239 *Moment-based analyses of historical population changes*

240 Historical population fluctuations were inferred using three moment-based methods. First, the  
241 mismatch distribution of pairwise differences between  $NC_{total}$  haplotypes was plotted. The  
242 observed distribution of pairwise differences was compared to the expected number of pairwise  
243 differences under a model of population expansion with 1,000 bootstrap replicates in Arlequin v.  
244 3.1 (Excoffier, Laval & Schneider, 2005). Second, Harpending's raggedness index ( $r$ ) was used  
245 to test whether the observed distribution was significantly different from the expected theoretical  
246 distribution under a model of expansion (Harpending, 1994). Third, an alternative metric, Fu's  $F_s$   
247 statistic, was used to test the selective neutrality of mutations (Fu, 1997). Fu's  $F_s$  can detect an  
248 excess (or deficiency) of haplotypes, given the observed haplotype diversity, thereby indicating a  
249 population expansion (or contraction).

250

251 *Coalescent analyses of historical population changes*

252 Two Bayesian coalescent MCMC models were used to estimate historical demographic  
253 fluctuations of *P. marinus* in Lake Champlain over time. First, the program BEAST v.1.6.2  
254 (Drummond & Rambaut, 2007) was used to make inferences based on NCII sequences. This  
255 region was selected because of its high concentration of polymorphic sites. BEAST applies a  
256 Bayesian coalescent-based procedure, using MCMC to sample the posterior distribution of  
257 genealogical trees, demographic parameters over time, and coalescent events given sequence  
258 information and a set of priors. For our demographic model, we applied the Bayesian Skyline  
259 Plot (BSP). BSP is a change-point model that, assuming a single panmictic population, estimates  
260 fluxes in population size through time and uses a smoothing procedure to visualize these changes

261 (Drummond et al., 2005). To determine which nucleotide substitution model fit the data, Akaike  
262 information criteria (AIC) values were calculated in jMODELTEST v.0.1.1 (Guindon &  
263 Gascuel, 2003). The substitution model selected by AIC was then used as a prior in BEAST. For  
264 our baseline study, we assumed a strict molecular clock of  $3.6 \times 10^{-8}$  substitutions per base per  
265 year (i.e., 3.6% substitutions per million years), based on previous estimates of divergence in  
266 mtDNA noncoding regions in fishes (Donaldson & Wilson, 1999). The BSP group number ( $m$ )  
267 was set to 15. The parameter  $m$  allows adjacent coalescent intervals to be grouped so that they  
268 can have the same  $N_e$ —it serves to smooth the resulting BSP. The maximum effective population  
269 size was set at 10,000, a high estimate based on preliminary coalescent model runs. MCMC  
270 chains were run for  $50^6$  iterations in triplicate, sampling the posterior distribution every 1,000  
271 iterations. The sampling distribution of the model was evaluated in TRACER v.1.5, with the first  
272 10% discarded as burn-in (Rambaut et al., 2014). Quality of the MCMC convergence was  
273 assessed by the effective sample sizes (ESS): if the ESS value was less than 100, it was assumed  
274 that the MCMC chain had not been run long enough to get an accurate representation of the  
275 posterior distribution and the trace was discarded (Drummond et al., 2007).

276         A sensitivity analysis for the BEAST modeling was carried out on two parameters—clock  
277 rate and maximum effective population size—to assess whether priors biased parameter  
278 estimates. These parameters were selected because the priors were based on our preliminary  
279 estimates. For each parameter change, three additional MCMC chains were run keeping all other  
280 baseline input values constant. We used two alternative maximum population sizes (20,000 and  
281 100,000) and at one alternate clock rate ( $2.0 \times 10^{-8}$  substitutions per base per year, i.e. 2.0%  
282 substitutions per million years). To determine how sensitive model results were to the priors,  
283 runs with altered priors were compared.

284 Historical demographic changes were also inferred from coalescent modeling of nuclear  
285 microsatellite data. Allele frequencies at eight microsatellite loci, previously published by Bryan  
286 et al. (2005), were used as input for the Bayesian MCMC model Msvar v.1.3 (Beaumont, 1999).  
287 Msvar uses probable genealogies of allele frequency data to generate posterior probability  
288 distributions of four demographic parameters: current effective population size ( $N_0$ ), historical  
289 effective population size ( $N_1$ ), mutation rate ( $\mu$ ), and time since the demographic change began  
290 ( $t$ ) (Storz & Beaumont, 2002; Beaumont & Rannala, 2004). Broad prior distributions were  
291 defined for each parameter to test whether the model could detect true population fluctuations  
292 (Table S1). Five independent chains were run for a panmictic sea lamprey population (Bryan et  
293 al., 2005; Waldman, Grunwald & Wirgin, 2006), under a model of exponential growth, with an  
294 average generation time of six years (Hardisty & Potter, 1971), and a  $N_e/N_c$  ratio of 0.2  
295 (Frankham, 1995). Each chain consisted of  $8 \times 10^8$  iterations, with sample points taken once  
296 every 10,000 iterations. The burn-in portion of the chain was excluded by truncating runs to  
297 consider only the second half. To estimate each demographic parameter, the five truncated  
298 chains were combined into one posterior distribution and peak density values were recorded for  
299 each parameter, along with 95% highest probability density (HPD) credibility intervals. A Bayes  
300 factor was also calculated to test whether there was more support for a population expansion or  
301 contraction using the method described by Storz and Beaumont (2002). Under the Bayesian  
302 statistical framework, a Bayes factor is an appropriate metric for comparing alternative models.

303 Sensitivity analyses were not conducted for Msvar because the model is computationally  
304 demanding; however, simulation modeling has revealed that Msvar has the power to detect true  
305 population contractions even with broad priors (Girod et al., 2011). These simulation models  
306 have also shown that the precision of estimates increases when demographic parameters are

307 scaled using coalescent theory. Thus the marginal posterior distributions of the following scaled  
308 parameters were also plotted:  $\theta_0 = 4N_0\mu$  and  $\theta_1 = 4N_1\mu$  (effective population sizes scaled by  
309 mutation rate);  $t_f = t/(2N_0)$  (time scaled by current effective population size).

310

## 311 RESULTS

### 312 *Haplotype diversity & population structure*

313 We observed 14 unique haplotypes among the 54 NC<sub>Total</sub> mitochondrial sequences (Table 1). The  
314 haplotype network of all 14 NC<sub>total</sub> haplotypes visualizes their relationships as well as their  
315 relative abundance (Fig. 1). Haplotype 1 is predominant in Lake Champlain, with five other rare  
316 haplotypes (2, 3, 6, 7, and 9) having only one base pair difference from haplotype 1. Most of the  
317 mutations within these five haplotypes are point mutations within NCI or the 5' end of NCII  
318 (Table 1). In contrast, most of the remaining rare haplotypes branch off of haplotype 7 and are  
319 characterized by mutations within the A/T-rich, repetitive 3' region of NCII. In general, these  
320 mutations in the repetitive region were retained in each subsequent repeat of the sequence,  
321 leading to a larger number of overall mutations in NCII.

322 An exact test of population differentiation revealed that haplotypes were randomly  
323 distributed across the three sampling locations (n=54; p=0.265; # dememorization steps =  
324 10,000; # steps Markov chain = 100,000). Thus, the Lake Champlain population is considered to  
325 be panmictic, consistent with results from previous studies (Bryan et al., 2005; Waldman,  
326 Grunwald & Wirgin, 2006).

327

328 *Moment-based analyses of historical population changes*

329 The mismatch distribution of  $NC_{total}$  was smooth and unimodal, suggesting that the sea lamprey  
330 population has undergone an expansion (Fig. 2). The distribution of observed pairwise  
331 differences closely matches the expected distribution of pairwise differences under a model of  
332 population expansion, with a raggedness index that was positive, but not significant ( $r = 0.04$ ;  $n$   
333  $= 54$ ;  $p = 0.86$ ). This indicates that there was no significant deviation from the theoretical model  
334 of expansion. Additionally, Fu's  $F_s$  statistic was significantly negative ( $F_s = -6.61$ ;  $n = 54$ ;  $p =$   
335  $0.02$ ), indicating an excess of rare haplotypes, which would be predicted under a scenario of a  
336 recent population expansion. Overall, these moment-based methods of studying historical  
337 demography with mtDNA data strongly support a population expansion, without explicit  
338 estimates of the timing and magnitude of the event.

339

#### 340 *Coalescent analyses of historical population changes*

341 For the BEAST analysis of mtDNA NCII data, the best-fit model of sequence evolution  
342 determined by jMODELTEST was Hasegawa, Kishino and Yano (HKY) + I, where I means  
343 there is a significant proportion of invariable sites. The BSP derived from the NCII data shows a  
344 decrease in effective population size starting around 400 years ago that continued until 50 to 100  
345 years ago, at which point effective population size slowly began to increase (Fig. 3). The  
346 sensitivity analysis showed that the maximum population size prior was directly proportional to  
347 the current  $N_e$  estimate; when doubled, the  $N_e$  estimates also doubled. However, changes in the  
348 maximum population size and clock rate priors did not affect the overall pattern of the BSP  
349 (Table S2).

350 In contrast to the BEAST analysis, coalescent modeling of microsatellite alleles estimated  
351 one historical population contraction. The five Msvar chains converged below a critical potential

352 scale reduction factor of 1.2 (Brooks & Gelman, 1998) for all four demographic parameters,  
353 indicating that there was good convergence of parameter estimates. The Bayes factor (BF) for a  
354 population contraction indicated very strong support for a population contraction over a  
355 population expansion ( $2 \cdot \ln(\text{BF}) = 10.3$ ) (Kass & Raftery, 1995), with the vast majority of  
356 MCMC iterations estimating a population contraction (Fig. 4a). The posterior density  
357 distributions reveal the estimates for each demographic parameter with 95% HPD intervals (Fig.  
358 4b-d). There was a decrease in effective population size from approximately 2,660 ( $\text{HPD}_{95\%} =$   
359 153, 65750) down to approximately 50 currently ( $\text{HPD}_{95\%} = 0.06, 1440$ ) (Fig. 4b). The density  
360 peak was higher, with narrower credibility intervals, for historical effective population size  
361 compared to current effective population size. This population contraction was estimated to have  
362 begun approximately 820 years ago, although the peak density was only 40% for this time  
363 parameter, with wide credibility intervals ( $\text{HPD}_{95\%} = 0.81, 71558$ ) (Fig. 4c). Finally, there was a  
364 high density peak for the mutation rate estimate, with over 80% of the estimates approaching a  
365 modal value of  $3.26 \times 10^{-4}$  mutations/site/generation ( $\text{HPD}_{95\%} = 3.52 \times 10^{-5}, 2.78 \times 10^{-3}$ ) (Fig.  
366 4d).

367 The posterior distributions of the scaled demographic parameters also indicate a  
368 population contraction (Fig. 5). As predicted by simulation analyses, scaling the parameters  
369 increased the precision of parameter estimates. The magnitude of the scaled population size  
370 estimates were very similar to the unscaled estimates. Modal  $\theta$  estimates corresponded with a  
371 decline from approximately 2,335 individuals historically ( $\text{HPD}_{95\%} = 359, 29614$ ) to 60  
372 individuals currently ( $\text{HPD}_{95\%} = 0.08, 772$ ). Despite increased precision, there was still some  
373 overlap in the 95% credibility intervals for  $\theta$  (Fig. 5A). In contrast, there was a substantial  
374 increase in precision of the time estimate, suggesting a more ancient demographic event: the

375 modal value of the scaled time distribution corresponded with an event beginning approximately  
376 1,230 years ago (HPD<sub>95%</sub> = 128, 7522) (Fig. 5B).

377

## 378 **DISCUSSION**

379 Genetic-based methods can be useful tools for studying demographic changes in the absence of  
380 historical population records, though their precision is linked to the timing and magnitude of the  
381 events themselves (Beaumont & Rannala, 2004; Lawton-Rauh, 2008). In this study, we applied  
382 multiple analytical approaches and used two sets of genetic markers to investigate the population  
383 history of *P. marinus* in Lake Champlain. Synthesizing the results to draw conclusions about the  
384 most likely historic scenario of demographic change requires a critical analysis of the results  
385 generated by each method, an exploration of sea lamprey ecology, and consideration of the  
386 history of land-use and management practices in the region.

387

### 388 *Moment-based analyses of historical population changes*

389 Mismatch distribution analysis is based on the assumption that demographic changes leave  
390 corresponding genetic signatures in neutral sequence data and gene trees (Rogers & Harpending,  
391 1992; Harpending et al., 1998). The mismatch distribution of this study fits a model of  
392 population expansion well, being both smooth and unimodal. However, a major weakness of this  
393 method is its poor ability to make specific inferences about the timing of demographic changes.  
394 In general, when using moment-based methods, only a general inference as to whether the  
395 change was relatively recent or ancient can be made. The presence of many low-frequency  
396 mutations is one indication of a 'recent' expansion (Schneider & Exoffier, 1999). This pattern  
397 was evident in the haplotype network, in which the majority of haplotypes in Lake Champlain

398 were present in only one to three individuals (Fig. 1). Additionally, the mean number of pairwise  
399 differences in the mismatch distribution can be used for a crude estimate of timing: a 'low' mean  
400 can indicate a 'recent' expansion while a 'high' mean can indicate a more 'ancient' expansion  
401 (Okello et al., 2005). The haplotypes in the Lake Champlain lamprey population had a mean of  
402 1.74 mismatches, which supports a 'recent' expansion. Yet, without explicit guidelines for  
403 differentiating between 'low' and 'high', or inferring what 'recent' and 'ancient' indicate on an  
404 evolutionary timescale, caution should be used in interpreting these results. Instead, these results  
405 are more appropriately used as supplemental evidence to be considered alongside inferences  
406 made by coalescent-based analyses.

407

#### 408 *Coalescent analyses of historical population changes*

409 BEAST and Msvar analyses both detected an initial decline in effective population size. These  
410 congruent results between two separate genomes strengthen the evidence for a decline; however,  
411 the magnitude and timing of the event(s) remain uncertain. The BSP generated using BEAST  
412 provides credibility intervals at every time point that reflect both coalescent and phylogenetic  
413 uncertainty. This proved to be important, as the BSP shows that the credibility intervals for  
414 effective population size ranged from 51 to 9,500. These wide credibility intervals around  $N_e$   
415 could be due to the limited number of segregating sites within the NCII sequence data. The  
416 sensitivity analysis also showed that the maximum population size prior affected the BSP results,  
417 with the estimated  $N_e$  being proportional to the prior. However, the overall pattern of  
418 demographic change was consistent regardless of the priors (see Table S2), indicating that the  
419 estimated mean  $N_e$  always showed a contraction followed by an expansion.

420 In contrast to BEAST, Msvar estimates four parameters simultaneously and can only  
421 detect the single most likely demographic trend over time based on the strongest genetic signal.  
422 Bayesian models, such as Msvar, require prior distributions for each parameter (Storz &  
423 Beaumont, 2002; Beaumont & Rannala, 2004). Generating priors can be difficult when no  
424 accurate estimates of demographic parameters are available, as is the case for sea lamprey in  
425 Lake Champlain. Estimates of current sea lamprey population size are poor because census data  
426 are inferred from wounding rates on prey species. Moreover, no historical documentation of sea  
427 lamprey population size in Lake Champlain exists. In the absence of reliable data, broad and  
428 equivalent priors were used for current and historical effective population size (Goossens et al.,  
429 2006). Furthermore, through simulation analyses, Girod et al. (2011) showed that Msvar is  
430 powerful at detecting population fluctuations even without informed priors if the demographic  
431 event was sufficiently large and ancient. The contraction detected in this analysis indicates that  
432 effective population size has declined from roughly 2,660 to 50 (a low but plausible estimate  
433 given the aggressive population control program and the fish's high fecundity). It is important to  
434 note that these values represent best estimates based on the mode of the posterior density  
435 distributions, but that the 95% credibility intervals of historical and current population size  
436 overlap due to the uncertainty associated with the current effective population size parameter.  
437 Scaling the population size parameters by the mutation rate did increase precision, but there was  
438 still overlap in the credibility intervals. Thus, for a comprehensive population history, the Msvar  
439 results must be interpreted alongside results from the BEAST and moment-based analyses.

440

441 *Scenario 1: Evidence for fluctuations within a native population*

442 Capitalizing on multiple analytical approaches and two unique genetic data sets used in this  
443 study, it is possible to infer multiple demographic fluctuations. While the mismatch distribution  
444 detected a population expansion and coalescent modeling using Msvar detected a decline, the  
445 results may not be in conflict. Instead, they may be detecting the two different demographic  
446 changes that are both evident in the BSP generated by BEAST. By considering known events in  
447 the ecological history of Lake Champlain and the approximate time scale estimated by each  
448 method, we present the most parsimonious demographic history of a native sea lamprey  
449 population in Lake Champlain.

450         The decline in effective population size detected by both coalescent models may correlate  
451 with land-use changes and fishing pressures that began after the arrival of European settlers in  
452 the mid-18<sup>th</sup> century. The BSP from BEAST indicated that the decline began approximately 400  
453 years ago. Given that there is uncertainty around this time estimate, one hypothesis is that a  
454 gradual decline in population size began around 1753 when a large number of European settlers  
455 arrived in the area. At this time, human impact on the landscape increased with the onset of  
456 large-scale agricultural and clear-cutting practices and, later, the construction of mills and dams  
457 that would have limited upstream lamprey spawning migrations (Klyza & Trombulak, 1999).  
458 Moreover, the acceleration of the decline shown in the BSP correlates with the extirpation of  
459 salmonids in Lake Champlain—the primary food source of parasitic-phase sea lamprey. The last  
460 documentation of native Atlantic salmon in the basin was 1830, with lake trout extirpation  
461 following shortly thereafter in the 1890s (Fisheries Technical Committee, 2009). Thus, it is  
462 plausible that the population contraction detected by both coalescent models can be explained by  
463 known ecological changes in the region, though some caution is warranted given that the scaled  
464 Msvar distributions suggest that the decline was more ancient (ca. 1230 years ago).

465           The mismatch distribution analysis may be detecting a more recent population expansion  
466 associated with salmonid stocking in Lake Champlain. This recent expansion is supported by the  
467 upward trend in effective population detected in the BSP in the last 50 years. A stocking program  
468 began for both Atlantic salmon and lake trout in 1972 and now focuses on stocking the main part  
469 of the lake with yearlings (Marsden et al., 2003; Fisheries Technical Committee, 2009). The  
470 reintroduction of their primary food source could have allowed the sea lamprey population to  
471 expand.

472           While this demographic history is plausible given the ecological context of Lake  
473 Champlain over the last 300 years, it is important to consider whether the data represent two real  
474 population changes, or whether these changes are artifacts of the models. To address this issue,  
475 an important consideration is the difference in mutation rates between the nuclear and  
476 mitochondrial markers used in the study, as faster-evolving markers generally have a greater  
477 power to detect more recent events. While the mitochondrial genome tends to have a higher  
478 mutation rate than the nuclear genome due to the mutagenic properties of respiration by-products  
479 and the limited DNA repair mechanisms of the mitochondrial genome (White et al., 2008),  
480 nuclear microsatellites have an even higher average mutation rate due to strand slippage. Indeed,  
481 Msva estimated a fast mutation rate for the microsatellites used in the study, with the narrow  
482 posterior distribution peaking at  $3.24 \times 10^{-4}$  mutations/site/generation. Nevertheless, Msva, a  
483 model that uses microsatellite markers exclusively, has been shown to have a bias towards  
484 detecting ancient declines in effective population size (Beaumont, 1999). Most recently, this  
485 tendency has been rigorously confirmed through simulation studies (Girod et al., 2011).  
486 Therefore, even if there were signatures of two demographic fluctuations in the population's  
487 history, Msva, which can only identify a single event, is predicted to detect the more ancient

488 decline. As to whether or not the mtDNA sequence data show a true signature of expansion  
489 within the last 40 years, some insight can be gained from a comparison between BEAST model  
490 inputs and outputs. Even though an extremely strict (i.e., slow) molecular clock was used as an  
491 input in this analysis, the posterior distribution of effective population size still shows an upward  
492 trend beginning approximately 50 years ago. Posterior estimates that deviate from prior  
493 predictions generally indicate a strong genetic signal. Taken together, these lines of evidence  
494 suggest that the population fluctuations detected are likely to be real.

495

496 *Scenario 2: Evidence for a founder event in the early 20<sup>th</sup> century*

497 Differentiating between founder events resulting from an invasion versus bottlenecks within  
498 native populations can be challenging, as both events are predicted to lead to a decline in  
499 effective population size and a loss of genetic diversity (Nei, Maruyama & Chakraborty, 1975).  
500 Thus, an alternative interpretation of the data is that the population contraction detected in Msvar  
501 represents a bottleneck immediately following an invasion from the anadromous Atlantic Coast  
502 population sometime after the completion of the Champlain Barge Canal in 1916, while the  
503 expansion detected in the moment-based and BEAST analyses represents the population boom in  
504 the 1970s. While the timing of such an invasion falls within the bounds of the BEAST and  
505 unscaled Msvar credibility intervals, both coalescent models assume a single panmictic  
506 population. Therefore, if the present-day Lake Champlain population is significantly  
507 differentiated from the anadromous population, the coalescent models could not accurately  
508 estimate historical changes beginning in the differentiated population of origin. Indeed, Bryan et  
509 al. (2005) found evidence for significant genetic structure between Lake Champlain and

510 anadromous sea lamprey populations, and long-term vicariance for the Lake Champlain  
511 population.

512         The mitochondrial haplotype network can provide additional insight into founder events:  
513 the random sampling associated with a *recent* founder event should lead to the loss of rare  
514 haplotypes. Therefore, a recent founder event is unlikely given the observed network, which  
515 features a number of rare haplotypes branching off of a predominant ancestral haplotype (Fig. 1).  
516 In sum, the majority of the evidence from the genetic models suggests that the population is  
517 native to Lake Champlain, but uncertainty remains and a founder event associated with an  
518 invasion in the early 1900s cannot be completely excluded given the genetic data that are  
519 presently available.

520

## 521 **CONCLUSIONS AND FUTURE DIRECTIONS**

522 This study builds upon previous research by modeling historical sea lamprey population  
523 fluctuations in Lake Champlain. Considering both coalescent models and moment-based genetic  
524 approaches, we conclude that multiple demographic events are likely to have occurred over the  
525 past 300 years. Importantly, however, there is a large amount of uncertainty around these  
526 estimates. While we argue that the data largely align with prior genetic studies and are most  
527 consistent with the native hypothesis, the wide credibility intervals around our estimates cannot  
528 exclude an alternative interpretation that a founder event occurred in the early 20<sup>th</sup> century.

529         As such, we propose two potential lines of future research aimed at resolving the  
530 residency debate. First, expanded genomic sampling could provide more accurate estimates of  
531 historical population sizes within Lake Champlain, as well as the timing of divergence between  
532 the Lake Champlain and Atlantic populations. Second, given the uncertain results from neutral

533 genetic data, it would be useful to investigate genes that may be under selection, e.g., those  
534 regulating Na/K-ATPase pumps in the gills. Population genetic analyses, gene expression  
535 analyses, and physiological saltwater challenges could be undertaken in tandem. Together, these  
536 lines of research may provide further clarity to the history of sea lamprey in Lake Champlain.

537

#### 538 **ACKNOWLEDGEMENTS**

539 The authors thank U.S. Fish and Wildlife Service professionals in Essex Junction, VT for  
540 providing sea lamprey tissue. We also thank Vicenta Hudziak and Livingston Burgess for  
541 assistance in DNA sequencing and Jeremy Ward, Wayne Bouffard, Steve Smith, and Brad  
542 Young for providing feedback. We are grateful to John Waldman, Amy Russell, and one  
543 anonymous reviewer for helpful comments on the manuscript. The findings and conclusions in  
544 the article are those of the authors and do not necessarily represent the views of the USFWS.

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**Table 1** (on next page)

Polymorphic sites among 14 mtDNA haplotypes from the concatenated non-coding region sequences (NC<sub>total</sub>), relative to the most common haplotype (first entry).

Base pair positions relative to the reference mitochondrial genome are provided (Lee and Kocher, 1995). Dots indicate no change, dashes indicate a deletion, and G/C/A/T represent point mutations.

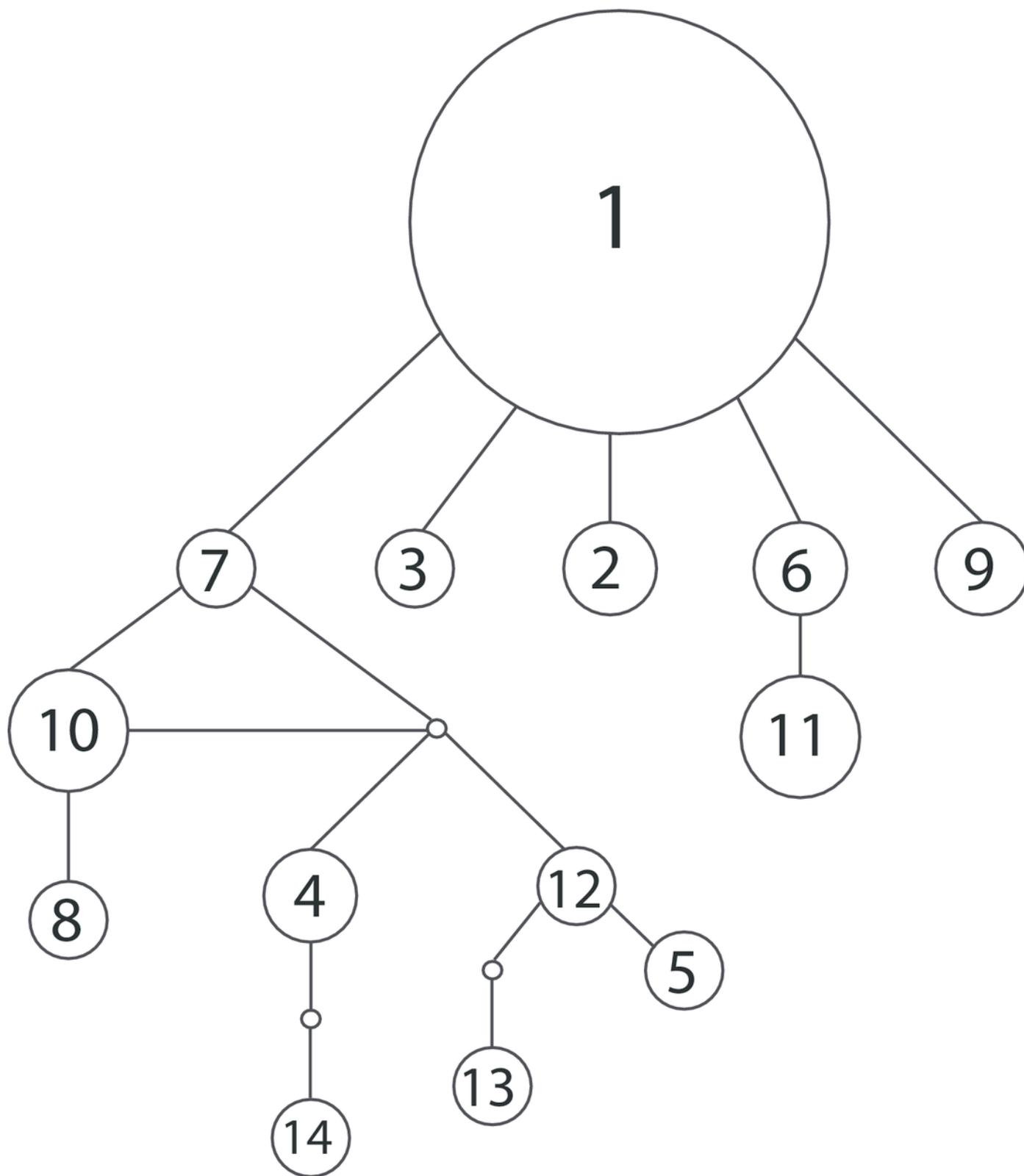
ID	Genbank Accession #	NCI		NCII									
		15398	15402- 15403	15403	16049- 16050	16062	16112	16114	16115	16134	16139	16142	16169
1	GU459340	G	-	C	T	T	T	T	C	T	T	C	C
2	GU459341	.	C	.	.	.	.	.	.	.	.	.	.
3	GU459342	.	-	.	.	-	.	.	.	.	.	.	.
4	GU459343	.	-	.	.	.	.	.	T	.	.	T	T
5	GU459344	.	-	.	.	.	A	.	-	.	.	T	T
6	GU459345	.	-	.	-	.	.	.	.	.	.	.	.
7	GU459346	.	-	.	.	.	.	.	.	.	.	.	T
8	GU459347	.	-	.	.	.	.	.	.	A	.	-	T
9	GU459348	A	-	.	.	.	.	.	.	.	.	.	.
10	GU459349	.	-	.	.	.	.	.	.	.	.	-	T
11	GU459350	.	-	-	-	.	.	.	.	.	.	.	.
12	GU459351	.	-	.	.	.	.	.	-	.	.	T	T
13	GU459352	.	-	.	.	.	.	-	-	.	A	T	T
14	GU459353	.	-	.	.	.	.	.	T	.	A	T	.

1

# 1

Haplotype network for 14 concatenated mtDNA haplotypes found in Lake Champlain, constructed using TCS v1.2.1 with 95% parsimony.

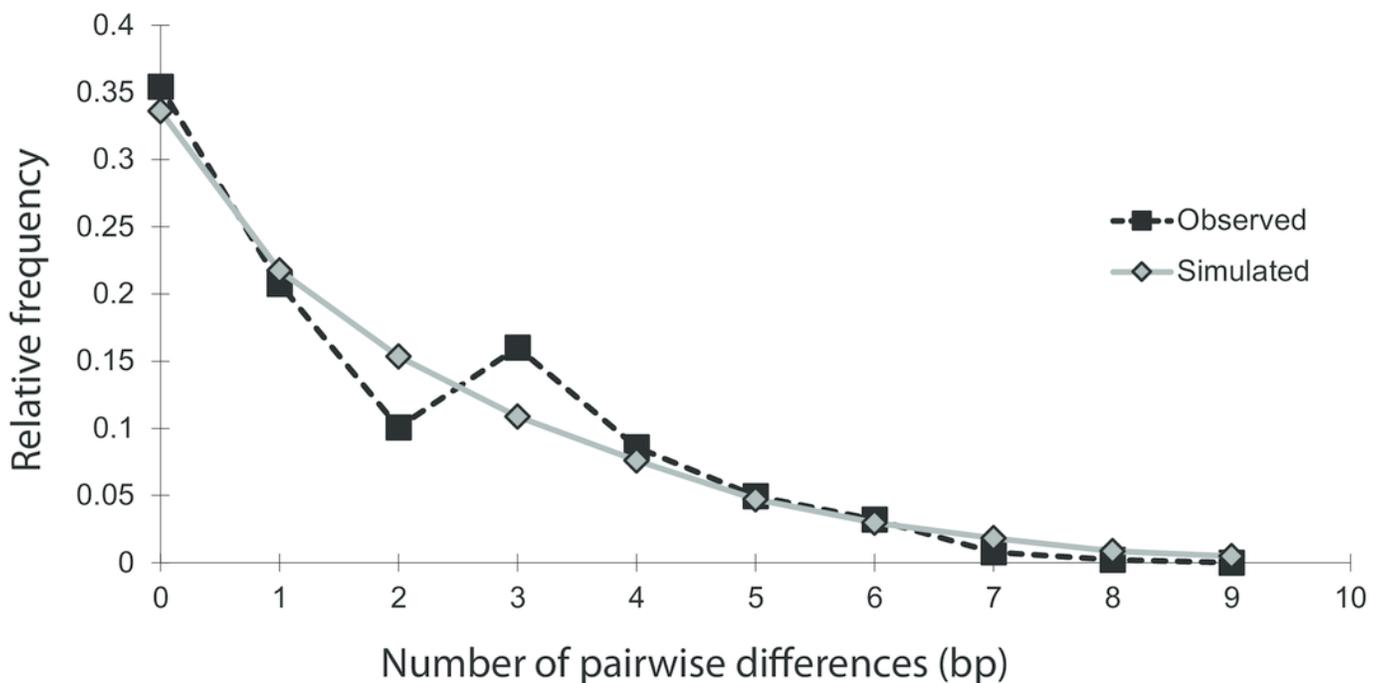
The size of the circle is proportional to the relative abundance of the haplotype. Solid lines represent one point mutation and small, unfilled circles represent inferred haplotypes. The mutations resulting in branching off of haplotype 7 are concentrated in the repetitive region of the NCII 3' region, which has an elevated mutation rate due to strand slippage.



## 2

Mismatch distribution of the 14 concatenated mtDNA haplotypes, conducted in Arlequin v.3.1.

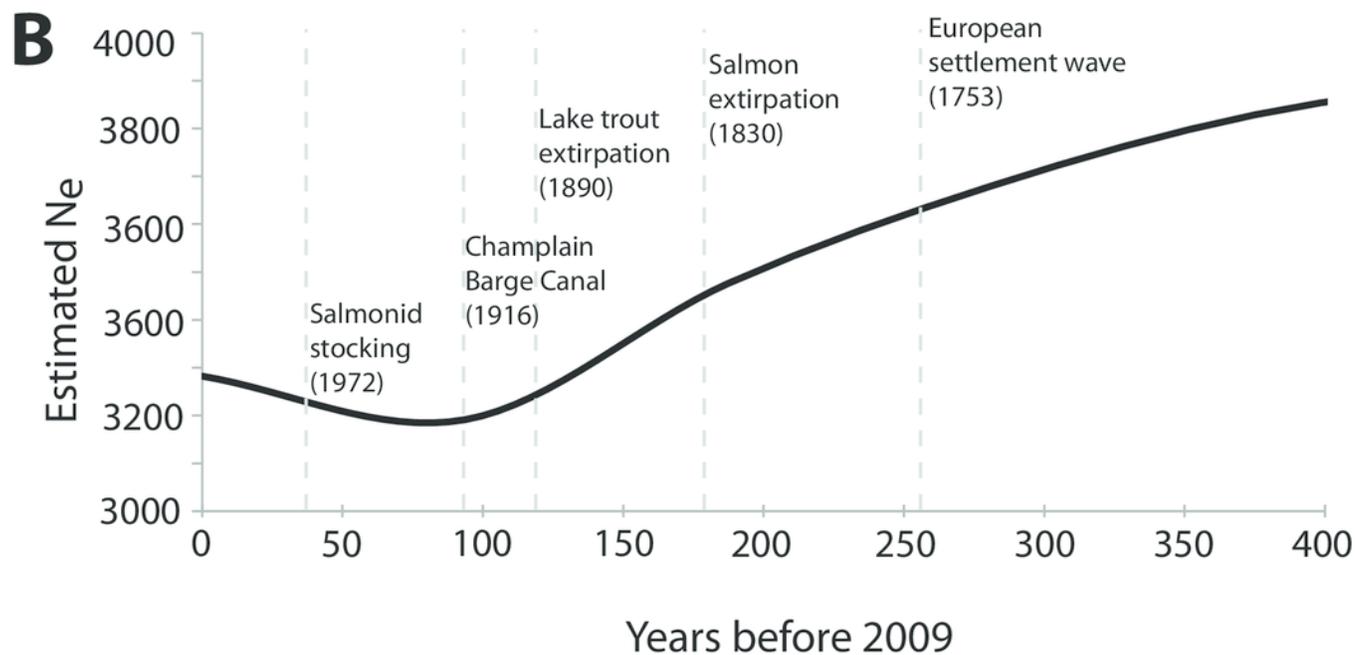
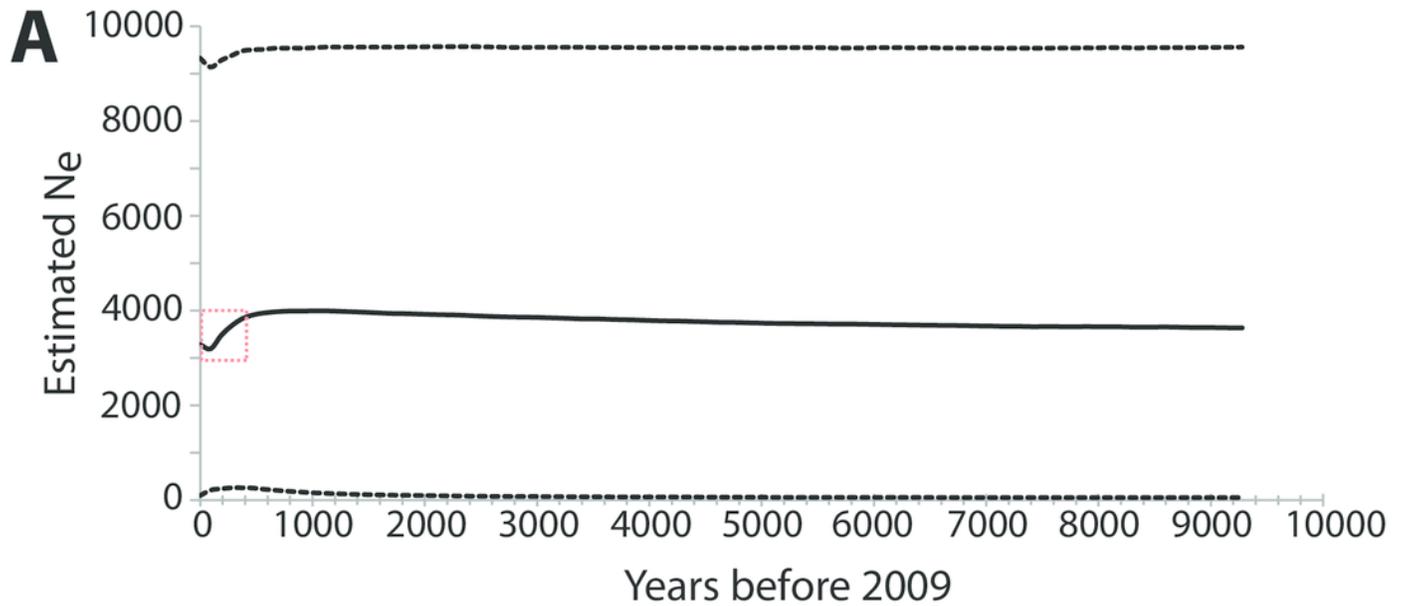
The black squares represent the observed number of pairwise bp differences between haplotypes, while the gray diamonds represent the expected number of pairwise bp differences between haplotypes, based on a model of population expansion (bootstrap replicates = 1,000). The relatively smooth and unimodal shape of the observed distribution closely matches the expected distribution for a demographic expansion.



## 3

Bayesian skyline plot (BSP) derived from NCII sequence alignments from Lake Champlain lamprey collected in May-June 2009.

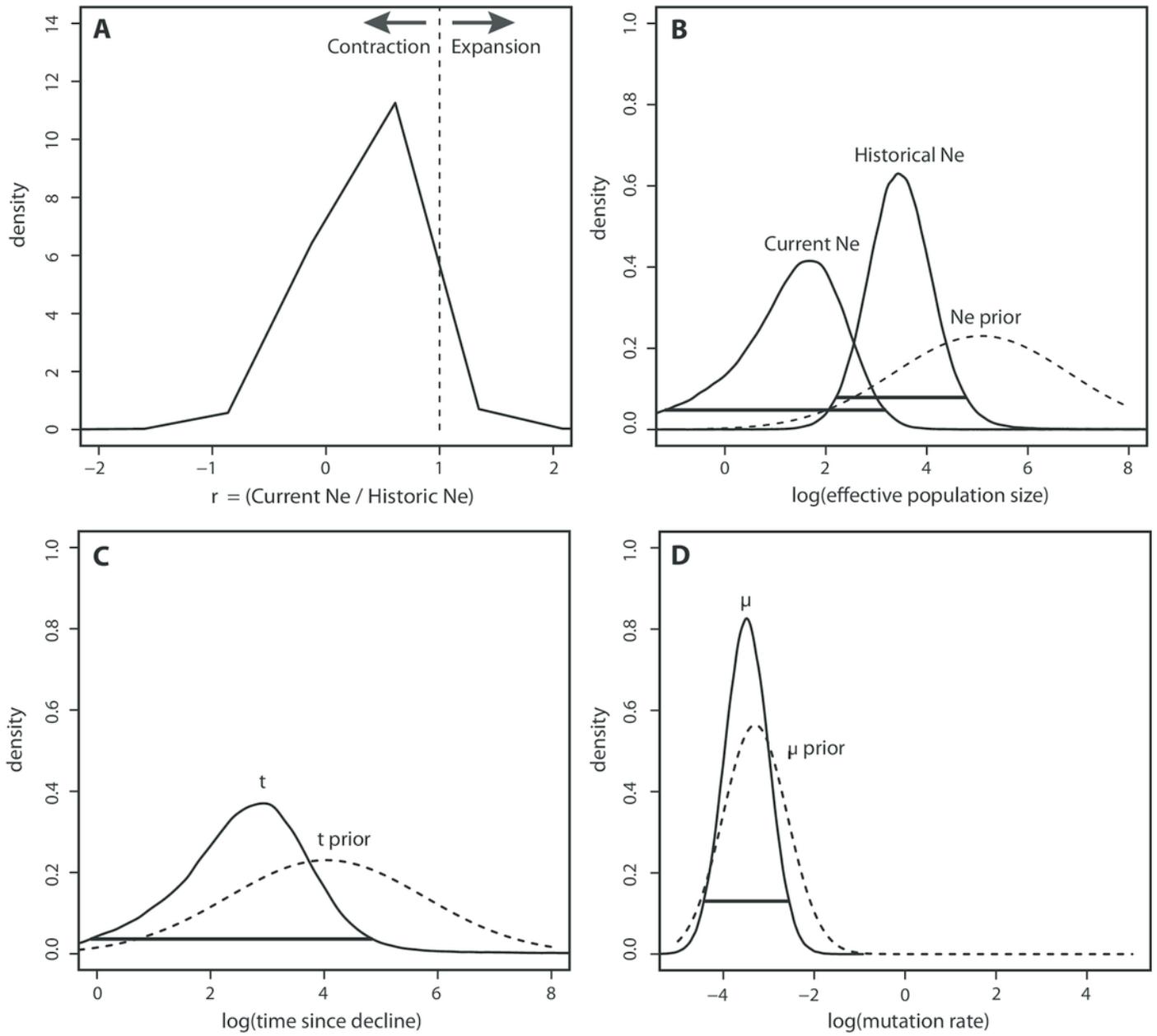
A) The mean effective population size is estimated from the Bayesian posterior distribution and is shown as the thick solid line. The horizontal dashed lines show the 95% HPD intervals around the  $N_e$  estimate. The dashed red box highlights the portion of the BSP shown in panel B; B) Zoom-in of the BSP for 400 years prior to 2009. The dashed vertical gray lines represent the timing of relevant historical events in the region. Note that the x and y axes differ between the two panels.



## 4

Msvar results based on microsatellites.

A) Density plot across MCMC iterations for  $r$ , which represents the ratio of current  $N_e$  to historical  $N_e$ , with  $r > 1$  indicating a population expansion and  $r < 1$  indicating a population contraction. B—D) Posterior density distributions for each demographic parameter, plotted on a log10 scale. Solid lines represent posterior density distributions and dashed lines represent prior distributions (with solid horizontal lines representing the 95% HPD credibility intervals for the posterior distributions); B) Effective population sizes; C) Time in years since population contraction; D) Mutation rate as # mutations/site/generation.



## 5

Marginal posterior distributions of scaled parameters from Msvar.

Solid lines represent posterior density distributions and dashed lines represent prior distributions (with solid horizontal lines representing the 95% HPD credibility intervals for the scaled posterior distributions). A) Scaled effective population sizes on a  $\log_{10}$  scale; B) Scaled time on a  $\log_{10}$  scale.

