Conservation Genomics of Urban Populations of Streamside Salamander (Ambystoma

barbouri)

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Abstract

In Tennessee, populations of the state endangered Streamside Salamander (Ambystoma 2 *barbouri*) are in decline as their distribution lies mostly within rapidly developing areas in the 3 Nashville Basin. Information regarding the partitioning of genetic variation among populations 4 5 of A. barbouri, and the taxonomic status of these populations relative to northern populations and their congener, the smallmouth salamander (A. texanum), have important implications for 6 management and conservation of this species. Here we combined mitochondrial sequencing and 7 8 genome-wide single nucleotide polymorphism (SNP) data generated using Genotyping-by-Sequencing (GBS) to investigate patterns of genetic variation within Tennessee populations of 9 A. barbouri, to assess their relationship to populations in Kentucky, and to examine their 10 11 phylogenetic relationship to the closely related A. texanum. Results from phylogenetic reconstructions reveal a complex history of Tennessee A. barbouri populations with regards to 12 northern populations, unisexual A. barbouri, and A. texanum. Patterns of mitochondrial 13 sequence variation suggest that A. barbouri may have originated within Tennessee and expanded 14 north multiple times into Kentucky, Ohio, Indiana and West Virginia. Phylogenetic 15 16 reconstructions based on genome-wide SNP data contradict results based on mitochondrial DNA and correspond to geographic and taxonomic boundaries. Variation in allele frequencies at SNP 17 genotypes, as identified by multivariate analyses and Bayesian assignment tests, identified three 18 19 evolutionary significant units (ESUs) for A. barbouri within the state of Tennessee. Collectively, these results emphasize the need for prioritizing conservation needs of Tennessee populations of 20 A. barbouri to ensure the long-term persistence of this species. 21

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Introduction

Genetic variation, population structure and demographic history are increasingly 24 recognized as important factors for the design of effective conservation strategies (Geist 2010). 25 Streamside salamander (Ambystoma barbouri, Fig. 1) populations in Middle Tennessee are 26 27 declining due to rapid urbanization in and around the Nashville Basin and as a result, were reclassified from "deemed in need of management" to state endangered by the Tennessee 28 Wildlife Resources Agency (TWRA) (Withers et al. 2009, TWRA 2018, Anderson et al. 2014). 29 Population fragmentation and loss of genetic variation that inevitably accompany loss of habitat 30 31 threaten the long-term adaptive potential and persistence of A. barbouri, but can be mitigated by management actions aimed at maintaining genetic diversity. Information regarding A. barbouri's 32 taxonomy, population structure and patterns of genetic variation are currently needed to 33 prioritize conservation needs and to efficiently allocate management resources. Genomic tools 34 are increasingly being used to improve recovery and management planning in at-risk salamander 35 species by informing taxonomic relationships, demographic histories, and biologically 36 meaningful units of conservation that will preserve genetic diversity and the long-term adaptive 37 potential of species. 38

Taxonomic uncertainties surrounding the evolutionary relationship between *A. barbouri* and its sister-species *A. texanum* (smallmouthed salamander) are a concern for precisely identifying targets of conservation. These two species are nearly indistinguishable based on external morphological features alone and were previously considered to be conspecific. They can be differentiated using scanning electron microscopy on the number and shape of maxillary and premaxillary teeth, life-history traits, and choice of breeding environment (Niemiller et al. 2009, Kraus & Petranka 1989). In addition to their morphological ambiguity, phylogenetic

reconstructions using mitochondrial and nuclear sequence data also produced conflicting results 46 regarding the relationship between A. texanum and A. barbouri (Niedzwiecki 2005). 47 Reconstructions based on mitochondrial sequences show that A. texanum and A. barbouri are 48 not reciprocally monophyletic; A. texanum is recovered as a clade nested within A. barbouri 49 from Tennessee, suggesting that pond-breeding A. texanum were more recently derived from a 50 51 stream-breeding A. barbouri ancestor from Central Tennessee. However, reconstructions based on two nuclear gene sequences resulted in reciprocally monophyletic clades for the two species, 52 consistent with a history where A. barbouri and A. texanum are older independent lineages 53 54 derived from a shared common ancestor. Additional sequence data are needed to resolve unanswered phylogenetic questions as these earlier reconstructions were based on a limited 55 number of molecular markers (two nuclear and one mitochondrial gene) and representative 56 outgroups. 57

Mitochondrial evidence also informs on the origins of unisexual Ambystomatid 58 populations that are common in the Great Lakes region of North America. Unisexual 59 Ambystomatids exhibit a unique reproductive mode whereby male sperm activates egg 60 development, but contributes variable amounts of nuclear genetic material depending on 61 62 compatibility with unrelated cytoplasmic DNA (termed kleptogenesis). While the nuclear genome unisexuals is a mixture of different Ambystomatid species, all known mitochondrial 63 64 haplotypes, across their range, nest within A. barbouri. Mitochondrial haplotypes are most 65 closely related to mitochondrial haplotypes found south of the Ohio River, and southwest of the Kentucky river (Bogart et al. 2007); however, the relationship of Tennessee A. barbouri to 66 67 unisexual Ambystomatids has not been explored.

68	At a finer scale, uncertainties also exist regarding the relationship among isolated
69	populations of A. barbouri in Tennessee and discontinuous populations in Kentucky, Indiana,
70	Ohio, and West Virginia. Phylogenetic reconstructions of A. barbouri populations based on
71	mitochondrial sequence data (913 bp) from the D-loop and an adjacent intergenic spacer suggest
72	that Tennessee A. barbouri populations are both ecologically and genetically distinct from more
73	northern populations (Eastman et al. 2009). At hatching, A. barbouri from Tennessee were
74	smaller and less developed than individuals from Kentucky (Niedzwiecki 2005). Also,
75	laboratory behavioral assays show that Tennessee A. barbouri were similar to western Kentucky
76	individuals, but less active than individuals from northern populations (Niedzwiecki 2005).
77	These differences were supported by genetic data; phylogenetic reconstructions recovered
78	mitochondrial haplotypes from Tennessee as monophyletic and basal to haplotypes from
79	Kentucky, Ohio, and West Virginia (Eastman et al. 2009). These results support an early
80	divergence of Tennessee populations and raise questions regarding the geographic origin of A.
81	barbouri. However, conclusions from this study were limited as these reconstructions are based
82	on a single mitochondrial gene from only three individuals from a few populations within a
83	kilometer of each other in Rutherford County.

Observations from field surveys suggest that *A. barbouri* populations in Tennessee are in decline, and the accompanying loss of genetic variation in small populations further threatens the long-term persistence of this species (Niemiller et al. 2006). Maintenance of genetic variation is a fundamental priority for conservation planning and requires information regarding the partitioning of genetic variation within and between isolated populations. Data on patterns of genetic variation in Tennessee populations of *A. barbouri* are limited to a handful of mitochondrial and nuclear sequences used for phylogenetic studies (Eastman et al. 2009;

Niedzweicki 2005). Genome-wide surveys of genetic variation from fine-scale population 91 sampling across the state are critical for assessing genetic variation within populations, 92 establishing units of conservation and maintaining historical patterns of gene flow between 93 populations to improve the outcome of recovery efforts (Shaffer et al. 1996). 94 Here we investigate patterns of genetic variation within and among populations of 95 96 Tennessee A. barbouri as well as the taxonomic relationship between A. barbouri and A. texanum in order to prioritize conservation needs and inform management practices aimed at 97 maximizing long-term persistence of this species. Mitochondrial sequence data and genome 98 99 wide SNP genotypes were used to (1) investigate the phylogenetic relationship between A. *barbouri* and *A. texanum*, (2) review the taxonomic relationship of disjunct populations of *A.* 100 *barbouri* in Tennessee relative to northern populations, (3) evaluate patterns of genetic 101 102 differentiation between geographically isolated populations of A. barbouri within the state of Tennessee and (4) estimate within-population genetic variation and examine demographic 103 history of Tennessee A. barbouri. Taxonomic relationships, units of conservation, and 104 geographic partitioning of genetic variation are discussed in the context of establishing 105 conservation priorities and designing effective management strategies for this species. 106 **Methods** 107 Tissue Collection and DNA Extraction. 108

Tissue samples of *A. barbouri* in the form of adult tail clips, eggs, and whole larvae were obtained from field surveys or from collaborators (Fig. 2, Table 1). Historical and predicted sites of *A. barbouri* were surveyed from January 2018 to March 2018 and again from November 2018 to March 2019, coinciding with timing of oviposition as reported by Niemiller et al. (2009). At each site, *A. barbouri* adults and eggs were collected by turning over cover objects within and

114	near seasonal streams. Additionally, pools were searched for free-swimming larvae; only one
115	larva was sampled from any single pool to avoid sampling related individuals from the same
116	clutch. A total of 235 individuals were included for genetic analysis. Samples included 225
117	individuals from 13 populations of A. barbouri in Tennessee spanning six counties in the
118	Nashville Basin: Bedford County (B6), Davidson County (D3), Rutherford County (R1, R7, and
119	R9), Sumner County (S2, S5, S7, S8), Wilson County (W1, W3, and W4), and Williamson
120	County (Wil2). A total of five A. texanum individuals were sampled; these included four
121	individuals from (Arnold Airforce Base, AAFB) and one from Craighead Co., Arkansas (AKT1).
122	Outgroups included two A. mabeei (North Carolina), two A. maculatum (AAFB), and one A.
123	talpoideum (AAFB).
124	DNA was extracted from tail clips, eggs, and larvae using the EZNA Tissue DNA Mini
125	kit (OMEGA BIO-TEK) following the manufacturer's protocol, except that DNA was eluted in
126	water. Approximately 1200 bp of the mitochondrial D-loop was targeted for PCR amplification

using primers developed by Shaffer and McKnight (1996; Table 2). The mitochondrial D-loop

has been shown to be informative for evaluating population structure and species relationships

within the genus *Ambystoma* (Bogart et al. 2007; Charney et al. 2014; Church et al.; 2003;

130 Shaffer & McKnight 1996; Zamudio & Savage 2003). Conditions for polymerase chain reaction

131 (PCR) were as follows: initial denaturation step of 2 minutes at 95°C, followed by 35 cycles of

132 15s at 95°C, 15s at 53°C, and 90s at 72°C. This program ended with a final extension of 10 min.

133 at 72°C. Amplified PCR product was cleaned prior to cycle sequencing by exonuclease I/shrimp

alkaline phosphatase (New England Biolabs) and used for bi-directional Sanger sequencing on

an ABI 3730 automated sequencer (MCLAB). Sequence chromatograms were imported and

136 visualized using SEQUENCHER 5.2 (Gene Codes Corporation). Sequences were aligned using

137 ClustalW Multiple Alignment option (Thompson et al. 2003) as implemented in Bioedit (Hall138 1999).

139 *Phylogenetic Reconstructions* for Mitochondrial Haplotypes

Phylogenetic reconstructions were estimated for all unique mitochondrial D-loop 140 haplotypes and an additional 53 A. barbouri, 83 unisexual Ambystomatids and 27 A. texanum 141 142 sequences obtained from Genbank. Both maximum likelihood (ML) and Bayesian optimality criteria were used for phylogenetic analyses. Maximum-likelihood analyses were performed 143 using the software RAxML (Stamatakis 2014) on the CIPRES Science Gateway (Miller et al. 144 145 2010) under the GTR+G model. Nodal support was estimated using rapid bootstrapping (1000 replicates). Bayesian phylogenetic reconstructions were performed using MrBayes 3.2.1 146 (Huelsenbeck & Ronquist 2001) also on the CIPRES Science Gateway. The best model of 147 substitution was selected by Modeltest (Posada & Crandall 1998), implemented in MEGA X 148 (Kumar et al. 2016) using Bayesian information criterion (BIC). The Markov chain Monte Carlo 149 (MCMC) algorithm ran for 10,000,000 generations, sampling every 1,000 generations. Two 150 independent runs were performed and the resulting trees were combined after the deletion of a 151 burnin (25%). A majority-rule consensus tree was generated and nodal support was estimated by 152 153 posterior probabilities.

154 *GBS Library Preparation and Sequencing.*

A total of 227 individuals were included for GBS sequencing including 201 individuals from 12 populations of *A. barbouri* in Tennessee, 17 individuals from five populations of *A. barbouri* in Kentucky, five individuals from two populations of *A. texanum*, and two sampled outgroups *A. talpoideum* (N=1) and *A. mabeei* (N=1; Table 1). Genotyping-by-Sequencing libraries were prepared using the restriction enzyme ApeKI following the protocol of Elshire et

160	al. (2011). Genomic DNA was quantified using Quant-iT Picogreen dsDNA Assay Kit (Thermo
161	Fisher Scientific), and all samples were standardized to between 5-6.5 ng/uL (50-65 ng total
162	genomic DNA). Extracted DNA was digested with the restriction enzyme ApeKI. Adaptors
163	containing PCR binding sites and individual barcodes were ligated onto digested DNA.
164	Barcoded DNA was pooled, and PCR was amplified using primers that bind to the ligated
165	adaptors (see Elshire et al. 2011 for primer sequences). The resulting PCR products were
166	cleaned using the Qiagen PCR purification kit and then cleaned again using the AxyPrep Mag
167	PCR Clean-up kit (Axygen, Big Flats, New York, USA). The distribution of the fragment size in
168	the PCR product was determined using an Agilent 2100 BioAnalyzer (Agilent Technologies
169	Inc., Santa Clara, CA, USA). Barcoded libraries were sequenced using the Illumina NextSeq
170	(Illumina Inc., San Diego, CA, USA) with a 75 bp single end read chemistry.
171	SNP Discovery and Filtering.
172	The Stacks program process_radtags was used to filter and demultiplex raw reads based
173	on barcoded sequences (Catchen 2011). We used the seven-step de novo clustering pipeline
174	ipyrad v. 3.5 (Eaton 2014) to generate and filter SNP datasets used in downstream analyses.
175	Quality filtering of raw sequence reads converted bases with Phred scores <33 to Ns, reads with
176	more than 5 Ns were removed. Reads were clustered using a sequence similarity threshold of
177	90% both within and between sampled individuals, with a minimum read depth of six.
178	Individuals with fewer than 500,000 reads were excluded from downstream analyses. Loci with
179	observed heterozygosity (H_o) greater than 0.5 were removed to filter out possible paralogs. The
180	final SNP dataset was then filtered to remove loci deviating from Hardy-Weinberg equilibrium
181	(p < 0.05), loci genotyped in less than 60% of individuals, and SNPs with a minor allele
182	frequency (MAF) less than 0.01. Only one SNP per tag was retained per locus.

183 *Population Genetic Variation and Effective Population Size*

184	Estimates of within population genetic variation for 12 Tennessee A. barbouri
185	populations were estimated using the R-package DiveRsity v. 1.9.9 (Keenan et al. 2013).
186	Summary statistics were estimated separately for each population and included the proportion of
187	SNPs that were polymorphic within each population (P), allelic richness (A_R), and observed and
188	expected heterozygosity (H_o and H_e). Estimates of P , A_R , H_o , and H_e were estimated for the
189	entire SNP dataset and for the subset of SNPs that were polymorphic within each population.
190	Effective population size (N _E) was estimated for Tennessee A. barbouri populations
191	using both the linkage disequilibrium (LD) method and the heterozygote-excess method
192	(Zhdanova & Pudovkin 2008) as implemented in the program N_E Estimator v2 (Do et al. 2014).
193	Populations with sample sizes of less than 20 individuals were excluded from analyses as these
194	datasets did not provide enough signal for reliable N_E estimation (Nunziata & Weisrock 2018).
195	The minor allele frequency (MAF) parameter was set at 0.01 and 95% confidence intervals were
196	estimated using the parametric chi-squared method.
197	Population Structure.
198	Population pairwise F _{ST} values were estimated for the all A. barbouri populations (Weir
199	& Cockerman 1984). Hierarchical partitioning of genetic variation across Tennessee populations
200	of A. barbouri was examined using Analysis of Molecular Variances (AMOVA) as implemented
201	by Arlequin 3.5 (Excoffier & Lischer 2010, Excoffier et al. 1992). Results from phylogenetic
202	reconstructions based on mitochondrial D-loop haplotypes were used to generate hypotheses

203 regarding higher-level structuring of populations. Significance of variance components was

204 determined using 1000 permutations.

205	The optimal number of genetic clusters (K) based on genomic SNPs was estimated using
206	both a multivariate approach and a Bayesian-based assignment method. Discriminant Analysis
207	of Principal Component (DAPC) was performed using the 'adegenet' package in R (Jombart &
208	Ahmed 2011, Jombart & Collins 2015). The find.clusters function was first used to identify the
209	optimal value of K based on a Bayesian Information Criterion (BIC) process. The optimal
210	number of principal components was determined using the function a.score. The optimal K was
211	then used to perform a DAPC analysis to describe the relationship between the genetic clusters
212	and individual membership probabilities were assessed using the function compoplot. DAPC
213	analysis was performed on three hierarchical datasets; these included 1) 12 populations of A.
214	barbouri in Tennessee, 2) all sampled Tennessee and Kentucky A. barbouri populations and 3)
215	all A. barbouri populations and four A. texanum individuals from AAFB.
216	Bayesian assignment tests were performed on Tennessee A. barbouri populations using
216 217	Bayesian assignment tests were performed on Tennessee <i>A. barbouri</i> populations using the program STRUCTURE 2.3.4 (Pritchard et al. 2003). Kentucky <i>A. barbouri</i> populations were
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217 218 219 220	the program STRUCTURE 2.3.4 (Pritchard et al. 2003). Kentucky <i>A. barbouri</i> populations were not included in assignment analyses due to small sample sizes at these sites. Values of <i>K</i> (number of populations) ranged from 1 to 12 populations with 20 replicate runs per value of <i>K</i> ; MCMC simulations were performed for a burn-in of 250,000 iterations and an additional
217 218 219 220 221	the program STRUCTURE 2.3.4 (Pritchard et al. 2003). Kentucky <i>A. barbouri</i> populations were not included in assignment analyses due to small sample sizes at these sites. Values of <i>K</i> (number of populations) ranged from 1 to 12 populations with 20 replicate runs per value of <i>K</i> ; MCMC simulations were performed for a burn-in of 250,000 iterations and an additional 1,000,000 iterations were retained for the final analysis. Results were summarized using the
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217 218 219 220 221 222 222	the program STRUCTURE 2.3.4 (Pritchard et al. 2003). Kentucky <i>A. barbouri</i> populations were not included in assignment analyses due to small sample sizes at these sites. Values of <i>K</i> (number of populations) ranged from 1 to 12 populations with 20 replicate runs per value of <i>K</i> ; MCMC simulations were performed for a burn-in of 250,000 iterations and an additional 1,000,000 iterations were retained for the final analysis. Results were summarized using the software package CLUMPAK (Kopelman et al. 2015). The optimal number of groups (<i>K</i>) was determined using the Delta-K method (Evanno et al. 2005) as implemented in STRUCTURE

A multispecies coalescent (MSC) model was used to examine taxonomic boundaries, 227 estimate species trees, and estimate population divergence times (t = years before present) as 228 implemented in the software BPP v. 4.4 (Yang 2015). Input datasets were formatted as full-229 length sequence alignments from our GBS sequencing and included all A. barbouri and A. 230 *texanum* populations. Loci represented by fewer than 100 individuals were removed from the 231 232 analysis. Joint species-delimitation and species-tree reconstructions were performed using the A11 model, where individuals were assigned to sampled populations. Run parameters were as 233 follows: uniform rooted trees were used as the species model prior, the theta prior ($\theta = 4N_E\mu$) 234 235 was assigned a gamma distribution with $\alpha=3$ and $\beta=0.04$ and the prior for tau ($\tau=t\mu$) was gamma distributed with α =3 and β =0.2. Three independent MCMC runs were implemented, each with a 236 237 burn-in of 5,000, a sample frequency of 10 and a total of 50,000 iterations. Species delimitation 238 and species-tree topologies with the highest posterior probabilities from the A11 runs were used as input for the A00 analyses. The A00 model estimates divergence time parameters and long-239 term N_E, assuming fixed species assignments and a fixed species tree. The priors for A00 were 240 the same as priors used for the A11 model. Three independent runs were also performed for the 241 A00 analysis, each with a burn-in of 5,000, a sample frequency of 10 and a total of 50,000 242 243 iterations. MCMC chains were pooled and absolute estimates of τ were converted to years before present (YBP) in the BPPR using the genome wide mutation rate estimate for the vertebrate 244 nuclear genome where μ = 1.21 x10⁻⁹ (Allio et al. 2019, Reis & Yang 2019). 245

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Results

247 *Mitochondrial Sequencing.*

Sanger sequencing of the mitochondrial D-loop resulted in an ~1100 bp sequence
alignment in the 81 individuals selected for sequencing. A total of 60 unique haplotypes were

identified from the newly sequenced individuals, 39 haplotypes were identified from *A. barbouri*populations sampled in Tennessee (See Supplement 1 for Genbank Accessions) and 15
haplotypes were identified from populations sampled in Kentucky. An additional six haplotypes
were sequenced from outgroups. The final alignment contained 126 variable sites (excluding
outgroups); of these, 120 sites were parsimony informative. No haplotypes were shared across
multiple populations.

256 *Phylogenetic Reconstructions.*

Both Bayesian and ML phylogenetic reconstructions of mitochondrial D-loop haplotypes 257 258 identified six major clades across A. barbouri and A. texanum haplotypes (Fig. 3). Clade I included all haplotypes belonging to the unisexual lineage of A. barbouri with 100% posterior 259 probability (pp) and 100% bootstrap support (bs) for Bayesian and ML analysis, respectively. In 260 261 *Clade II, A. barbouri* from Wilson Co. (populations W1, W3, and W4) and Davidson Co. (population D3) formed a monophyletic clade that also includes Kentucky A. barbouri 262 populations from southwest of the Kentucky River (pp = 100, bs = 95%). Clade I and Clade II 263 were recovered as sister clades in both analyses with strong support (pp = 100, bs = 99%). Clade 264 III included all A. texanum individuals (pp = 99, bs = 71) with the exception of a single A. 265 266 texanum haplotype obtained from Genbank (ID EU980569, collected from Lawrence KA); this sequence is recovered as basal to Clade III in both ML and Bayesian analyses. Clade IV groups 267 268 with *Clade III* and included Tennessee A. barbouri individuals from Rutherford Co. (populations 269 R1, R7, and R9), Bedford Co. (B6) and Williamson Co.; pp = 100, bs = 71%). Clade V included all Kentucky individuals from populations NE of the Kentucky River, Ohio and Indiana (pp =270 271 100, bs = 86%). Finally, *Clade VI* is sister to *Clade V* and included all Tennessee A. barbouri 272 from Sumner Co. (populations S2, S5, S7, and S8), a single haplotype from Wilson population

W3 and three haplotypes sequenced from Rutherford population R7 (pp = 100, bs = 99%). On a

274 larger scale, all *A. barbouri* and *A. texanum* are part of the same well-supported monophyletic

group in both analyses (pp = 100, bs = 100%), such that all *A. texanum* haplotypes are nested

within *A. barbouri*.

277 *GBS sequencing*.

The average number of retained sequence reads per individual generated from

sequencing of GBS libraries was 5,780,366; 220 individuals were retained after removal of low

coverage individuals (defined as < 500,000 total reads). A total of 1,558,844 loci were

recovered from the de novo assembly in ipyrad and 440,629 loci were retained after filtering.

Additional filtering for SNPs (i.e. filtering for low representation, HWE, MAF, and max

heterozygosity) recovered 1,169 SNPs for the dataset that included only Tennessee A. barbouri,

516 SNPs in the dataset that included all sampled *A. barbouri* populations in Tennessee and

Kentucky, and 500 SNPs in the dataset that included all *A. barbouri* and *A. texanum* populations.

286 Within Population Genetic Variation

Much of the genetic variation within our SNP dataset was in the form of fixed differences between populations (Table 3). The proportion of SNPs that were polymorphic within populations ranged from 0.045 (W3/Wil2) to 0.267 (S7). When only polymorphic loci were considered, the Will2 population had the highest estimates of Ho and He; however, this population had a low proportion of polymorphic SNPs overall. Rutherford County populations (R1/R7) had the lowest estimates of H_o and H_e, when considering only polymorphic loci and for the entire SNP dataset.

294 *Effective Population Size*

Estimates of effective population size (N_E) were generated for five populations with adequate sampling ($N \ge 20$), including S7, S8, R1, R7, and D3 (Table 4). Results from LD analyses ranged from $N_E = 15$ (R1) to $N_E = 108$ (108) (S8); the upper confidence interval for S8 was ∞ , indicating low signal for this dataset. Results from the heterozygote excess method were inconclusive (∞) for two of the five analyzed populations. Estimates from the remaining three populations ranged from 24 (D3) to 406 (S7); however, CIs included ∞ for two of these estimates (S7 and S8).

302 *Population Structure.*

Pairwise F_{ST} estimates for *A. barbouri* populations within Tennessee were geographically 303 structured in a hierarchical fashion (Table 5). Overall, F_{ST} values averaged 0.375 and ranged 304 from no differentiation ($F_{ST}=0$) to highly differentiated ($F_{ST}=0.725$, S5/R1). In general, 305 populations north of the Cumberland River (S2, S5, S7, and S8) showed evidence of long-term 306 isolation from populations south of the Cumberland. Pairwise F_{ST} estimates comparing 307 308 populations north and south of the Cumberland averaged 0.637, while F_{ST} estimates among Tennessee populations north of the Cumberland averaged only 0.043. Populations south of the 309 Cumberland were further structured; populations in Wilson and Davidson Counties (W1, W3, 310 311 W4, and D3) were diverged from populations further south in Williamson, Rutherford and Bedford counties. Estimates of pairwise F_{ST} comparing populations within Wilson/Davidson 312 313 averaged 0.121 and within Williamson, Rutherford and Bedford Counties averaged 0.070, while 314 between group pairwise F_{ST} values were much higher, averaging 0.300.

The three monophyletic clades of Tennessee *A. barbouri* recovered from mitochondrial phylogenetic reconstructions were further examined using genome-wide SNP genotypes in an AMOVA framework (Table 6). The majority of variance in SNP genotypes could be attributed

to differences between the three clades (63.64%). Differences between populations within clades
only accounted for 2.09% of the total molecular variance. Differences between individuals
accounted for the remaining 34.27% of variance in SNP genotypes.

Results of BIC for DAPC analysis suggested that three genetic clusters (K=3) best 321 represented genetic variation among populations sampled within Tennessee (Fig. 4A, Supp 2). 322 323 Membership of each cluster was geographically partitioned and consisted of a northern cluster that included populations from Sumner County, a central cluster comprised of individuals from 324 Wilson and Davidson Counties, and a southern cluster that represented Bedford, Rutherford, and 325 326 Williamson County populations. Membership probabilities and DAPC plots indicated some degree of admixture between the central and southern clusters, but no admixture within the 327 northern cluster (Supplement 2). A separate analysis that included Tennessee A. barbouri and 328 329 the five sampled populations from Kentucky also resulted in three genetic clusters (Fig 4B). Kentucky populations of A. barbouri were isolated from populations in Tennessee. Central and 330 southern populations were assigned to the same cluster and populations in Sumner county were 331 assigned to the third cluster. When A. texanum individuals were included in a third analysis, 332 results of the BIC analysis indicated the number of genetic clusters was five (Fig. 4C). 333 334 Tennessee populations were assigned to northern, central, and southern clusters and scatter plots showed Tennessee populations to be tightly grouped with the fourth cluster that included all 335 336 Kentucky populations. All A. texanum individuals formed the fifth cluster that was well 337 separated from all A. barbouri individuals.

Analysis of Bayesian assignment tests for Tennessee populations of *A. barbouri* were consistent with results from DAPC analysis that suggested the presence of three geographically partitioned genetic clusters within the state (Fig. 5). The optimal value of *K*, as determined by

the Delta K method, indicated four genetic clusters; however, increasing K from 2 to 3 did not
change population assignments. Assignment plots for K=2-3 separated northern populations in
Sumner County from central and southern populations. When K was increased to four,
populations from Davidson and Wilson Counties were separated from southern populations in
Bedford, Rutherford and Williamson Counties. Increasing the value of K to 5 increased the level
of admixture, but did not result in geographically meaningful partitions.

347 *Multispecies Coalescent Analyses.*

A total of 988 loci were retained for the MSC analyses performed by BPP. All three A11 348 runs recovered the same taxonomic groups and species tree topology with the highest posterior 349 probability. Out of 18 sampled populations, 14 populations were genetically distinct. Within 350 Tennessee, A. texanum and all A. barbouri populations, with the exception of R1/R7, were 351 352 identified as distinct groups (Fig. 6). Kentucky A. barbouri populations were split into two groups; populations FC, JC, RR, and SL were grouped together and the SW populations was its 353 own genetic group. The separation of individual populations in this analysis does not indicate 354 that each population warrants recognition at the species-level as simulations have shown that 355 BPP will split groups at the population-level when many loci are analyzed (Leach et al. 2019). 356 357 The A11 species tree reconstruction with the highest posterior identified the same Tennessee clades recovered in our mitochondrial gene trees with the exception of the placement 358 of A. texanum. In the species tree, A. texanum was basal to all populations of A. barbouri in 359 360 Tennessee and Kentucky. The estimated divergence time (A00 analysis) between A. texanum and A. barbouri was 1.6 million YBP. All Sumner A. barbouri formed a monophyletic group with 361 Kentucky populations as a monophyletic sister group. Divergence time estimates indicated that 362 363 Sumner populations and Kentucky populations shared a common ancestor < 500,000 YBP. All

A. barbouri populations south of the Cumberland River were monophyletic (TMRCA 260,000 364 YBP) and were further split into two groups that corresponded to clades recovered in the 365 mitochondrial gene tree. Davidson and Wilson populations were recovered together in the first 366 group (mitochondrial Clade II) and Rutherford, Bedford, and Williamson populations 367 (mitochondrial Clade IV) formed the second group. 368 369 Discussion Patterns of genomic variation and taxonomic relationships identified here have important 370 implications for developing management strategies aimed at the long-term conservation of A. 371 372 barbouri. Despite the documented decline of A. barbouri populations in the Nashville Basin, results from GBS-derived SNP genotyping indicate that estimates of genetic variation in extant 373 populations of Tennessee A. barbouri are similar to SNP-based datasets examined from other 374 ambystomatids. Partitioning of genetic variation between Tennessee populations suggests that 375 hydrogeography of the Nashville Basin has shaped patterns of gene flow. Both mtDNA and 376 genomic SNP genotypes showed similar patterns with respect to population structure within the 377 state of Tennessee that should be used to inform the designation of units of conservation. Our 378 results also reveal a complex history of Tennessee A. barbouri with more northern populations 379 380 and with its sister species A. texanum. Despite marked differences in phylogenetic reconstructions based on mtDNA sequencing and nuclear genotypes, collectively these results 381 382 indicate that Tennessee is genetically unique from Northern A. barbouri and A. texanum. 383 Genetic Variation within Populations. Loss of genetic variation resulting from rapid population decline threatens the long-term 384 success of conservation efforts due to loss of adaptive potential and fixation of deleterious alleles 385 386 (Kardos et al. 2021). For A. barbouri populations in the Nashville Basin, maintenance of

adaptive genetic variation is particularly critical as these populations are faced with habitat 387 alteration by urbanization in addition to climate change (Schmidt et al. 2021). Evaluating the 388 genetic health of populations is valuable for management and conservation planning; however, 389 the interpretation of genetic variation estimates from SNP data is challenging due to the limited 390 number of comparable studies utilizing reduced-representation methods in salamanders. 391 392 Heterozygosity estimates obtained here are similar to estimates obtained in the handful of published SNP-based genetic studies. In a recent survey of SNPs from ddRAD sequencing, 393 estimates of H_0 ranged from 0.165 to 0.269 for populations of the mole salamander (A. 394 395 *talpoideum*) and were slightly higher for populations of the marbled salamander (A. opacum; 0.207 - 0.298; Nunziata et al. 2017). Another SNP-based survey examined the effects of land 396 use on genetic variation in the northern two-lined salamander (*Eurycea bislineata*). Fusco et al. 397 (2020) reported nearly identical estimates of H_0 from urban, suburban and rural salamander 398 populations ($H_0 = 0.265$, 0.278 and 0.275, respectively) and concluded that genetic variation had 399 been maintained, despite habitat disturbance. However, estimates of heterozygosity may not be 400 directly comparable across independent studies as filtering parameters in bioinformatic pipelines 401 can influence population genetic summary statistics. Stringent filtering criteria may 402 403 preferentially retain loci in conserved regions of the genome and downwardly bias estimates of genetic variation (Huang & Knowles 2014). There have been previous population genetic studies 404 in A. texanum and A. barbouri populations based on microsatellite markers. Micheletti and 405 406 Storfer (2017) estimated genetic variation at 11 microsatellite loci in 76 populations of A. *barbouri* distributed throughout Kentucky, Ohio, and Indiana; the average H_e from their study 407 408 ranged from 0.67 - 0.81, respectively. Few conclusions can be made by comparing results from 409 these prior microsatellites studies to our results based on SNP genotypes; microsatellites are

A reduction in estimates of contemporary effective population size (N_E) may also signal a

410 known to have a much higher rate of mutation than nucleotide substitutions, increasing the

411 expected amount of allelic diversity for a given population size (Haasl & Payseur 2011).

412 *Effective population size.*

413

decline in the genetic health of at-risk populations. Our estimates of N_E were low (N_E from the 414 415 LD method averaged 58.0 across five populations), but were not unusual for LD-derived estimates of N_E in salamanders. Published estimates of N_E reported for amphibians have 416 typically been under 100 (Jehle & Arntzen 2002). Effective population size estimates reported 417 here were similar to values reported for the congeneric endangered California tiger salamander 418 (A. californiense; N_E values of 11-64; Wang et al. 2011), and for the long-toed salamander (A. 419 *macrodactylum*; $N_E = 23-207$; Funk et al. 1999). In *A. macrodactylum*, Savage et al. (2010) 420 421 estimated N_E for 47 breeding populations and more than half of these estimates were less than 50. Life history factors may provide some explanation for low N_E in this group as salamanders 422 often exhibit high variance in reproductive success and larval survival within populations. 423 Variance in reproductive success will increase relatedness among individuals in a population, 424 inflating linkage disequilibrium across loci and reducing N_E. 425 426 Obtaining robust estimates of N_E and other demographic parameters in at-risk species can be challenging due to the sample sizes required for LD detection. Many threatened taxa are 427 428 inherently rare and prohibitively difficult to sample in large numbers. For robust estimation of 429 N_E , it is recommended that sample sizes be greater than 30 for most systems (. When the number of sampled individuals is much smaller than the effective size, LD-based N_E estimates can be 430 431 downwardly biased and confidence intervals can be large (or infinite) due to inadequate signal in 432 the dataset (Waples & Do 2010). We limited our N_E analysis to populations with sample sizes >

433 20 individuals, which left only five populations for N_E estimation. The infinite upper confidence 434 interval for our N_E estimate in population Sumner 8 was likely due to an inadequate sample size

435 for this population (N=21).

436 *Phylogenetic Reconstructions and Population Structure*

Mitochondrial gene-tree reconstructions suggest a complex biogeographic history for A. 437 438 barbouri in Tennessee that involves populations of unisexual Ambystomatids, A. texanum and populations of A. barbouri from the northern core region (Kentucky, Indiana, and Ohio). Our 439 results support findings by Bogart et al. (2007) demonstrating a monophyletic relationship of 440 441 unisexual Ambystomatid mtDNA haplotypes nested within A. barbouri. The phylogenetic reconstruction shown here further details this relationship whereby unisexual Ambystomatids 442 likely shared a maternal common ancestor with A. barbouri populations in Kentucky and 443 populations from Middle Tennessee. This pattern, coupled with phylogenies based on nuclear 444 genes, points to a history of repeated hybridization events initiated by an A. barbouri maternal 445 446 ancestor in the southern portion of its range.

With regards to their mitochondrial lineage, A. texanum is nested within present-day A. 447 barbouri, rendering A. barbouri paraphyletic. This phylogenetic pattern contradicts the scenario 448 449 proposed by Kraus and Petranka (1989), that stream-dwelling A. barbouri descended from the more widespread pond-breeding A. texanum. Species-tree reconstructions based on genome-450 wide SNP data are incongruent with relationships recovered from the mitochondrial trees. This 451 452 species-tree topology adheres to conventional taxonomic and geographical boundaries (Fig. 6), where A. texanum is a sister-species to all A. barbouri populations. The TMRCA for A. barbouri 453 and A. texanum based on SNP genotypes was estimated at ~1.6 million YBP making this a 454 455 relatively recent split between these ecologically distinct species (Vences et al. 2007).

Incongruence between mitochondrial gene trees and reconstruction based on nuclear data are not
uncommon and are often attributed to ancestral lineage sorting, introgression, and/or sex-biased
dispersal (Nichols 2001, Toews & Brelsford 2012). Determining the cause of mitonuclear
discordance would require additional sampling across the distributions of both *A. barbouri* and *A. texanum;* however, it is relevant to note that recent studies have demonstrated that
hybridization is common across salamander lineages (including the genus *Ambystoma*) and may
facilitate rapid diversification (Patton et al. 2020).

The hydrogeography of the Central Basin, together with cyclic glacial movements during 463 the Pleistocene, may have shaped contemporary patterns of genetic variation in Tennessee 464 populations of A. barbouri. Mitochondrial-based phylogenetic reconstructions and partitioning 465 of genetic variation at SNP loci (including assignment tests, DAPC analyses, AMOVA, and 466 MSC reconstructions) identify the same three genetically distinct groups in Tennessee; these 467 include a northern cluster, a central cluster, and a southern cluster. The Cumberland River 468 appears to have served as a major barrier to gene flow between the northern and the 469 central/southern clusters. The central and southern clusters both occur south of the Cumberland 470 River and are divided by smaller regional drainage patterns. Individuals in the northern and 471 472 central clusters occupy the Old Hickory Lake and Lower Stones River watersheds, respectively. The central cluster is bordered by the Cumberland River to the North and by the East Fork of the 473 Stones River to the east. The southern cluster includes populations from three watersheds 474 475 including the West Fork Stones River Watershed, Upper Duck River Watershed, and Mill Creek Watershed; the southern cluster (with the exception of the Wil2 population) is bordered by the 476 477 Stones River to the north and by the Duck River to the south.

Mitochondrial-based phylogenetic reconstructions revealed that Tennessee A. barbouri 478 populations are not monophyletic with respect to northern populations, which may reflect a 479 history of repeated range expansions from Tennessee into the northern end of A. barbouri's 480 present day distribution. The estimated timing of divergence between the three Tennessee 481 clusters falls within the mid-late Pleistocene. Glacial advances during the Pleistocene did not 482 483 extend into Kentucky and Tennessee and this region likely served as a refuge for both terrestrial and aquatic fauna (Jacquemin 2016). As glaciers retreated, populations at the northern limits of 484 these refugia would have advanced northward into newly habitable territory. These expansions 485 486 would have left a genetic signature characterized by reduced genetic variation in the northern end of their distribution as a result of founder effects. Repeated expansions of A. barbouri 487 northward from Tennessee may explain the polyphyletic relationships of contemporary 488 Tennessee A. barbouri with respect to populations in the north. Specifically, mitochondrial A. 489 barbouri populations in northeastern Kentucky, Ohio and Indiana (Clade V) are nested within 490 Tennessee A. barbouri as are mitochondrial haplotypes from Ambystomatid unisexuals and A. 491 texanum. Genetic evidence of repeated northward expansions has been reported for the Eastern 492 Woodrat (Neotoma floridana; Hayes & Harrison 1992) and other southeastern fauna (see Hewitt 493 1995 for review). 494

495 Management Implications and Conclusions.

Results from our genomic survey have specific implications for the design of
management and conservation strategies that may improve the long-term persistence of *A*. *barbouri* in Tennessee. First, patterns of genetic variation in mitochondrial and nuclear genomes
support the assignment of three genetically distinct units for management that warrant the
designation of evolutionary significant units (ESUs), where ESUs are defined as groups of

501	populations that show phylogeographic differentiation for mtDNA haplotypes and divergence in
502	allele frequencies in nuclear markers (Moritz 1994, 1999). These three units include a Northern
503	ESU encompassing all Sumner County populations (S2, S5, S7 and S8), a Central ESU that
504	includes populations from eastern Davidson and Wilson Counties (D3, W1, W3 and W4), and a
505	Southern ESU that includes populations in Williamson, Rutherford, and Bedford Counties
506	(Wil2, B6, R1, R7, and R9). Patterns of differentiation between these three groups of
507	populations suggest a long history of genetic isolation at both mitochondrial and nuclear
508	markers, such that different groups are likely to possess unique combinations of adaptive genetic
509	variation and have likely experienced independent evolutionary trajectories.
510	Evolutionarily and phylogenetically distinctive populations contribute disproportionately
511	to genetic diversity and should be ranked highly in regards to conservation priority.
512	Geographically peripheral populations are frequently observed to be genetically less-redundant
513	than more central populations and are more likely to possess potentially adaptive genetic
514	variation (Volkmann et. al. 2014). For A. barbouri, Tennessee populations are geographically
515	disjunct from the majority of A. barbouri breeding populations (Niemiller et al. 2006) and
516	occupy the southernmost edge of the species distribution. Results from this study suggest that
517	Tennessee A. barbouri should be prioritized for conservation planning as these populations are
518	both genetically diverse and evolutionarily distinct from populations in the northern part of their
519	distribution. The distinctiveness of these populations is further evidenced by observed
520	differences in reproductive life-history traits, including mean diameter of early stage ovum size
521	and number of eggs per clutch (Niemiller et al. 2009). Prioritizing peripheral populations with
522	adaptive genetic variation and evolutionary potential is even more critical when considering
523	environmental challenges that accompany climate change. Amphibians in general are very

524	sensitive to climate change as their reproductive life histories are linked to temperature and
525	precipitation (Corn 2005). However, there is evidence that populations at the warm-range edge
526	of their distribution are more resilient (Razgour et al. 2019). Protection of A. barbouri breeding
527	sites in Tennessee may be instrumental to ensuring the long-term viability of this species as a
528	whole.
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534	

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Table 1. Population IDs, sample sizes for mtDNA/GBS analyses, and map coordinates for all populations.
Population IDs for *A. barbouri* collections correspond to map locations in Figure 1. For outgroup samples, AAFB denotes Arnold Air Force Base and AKT1 denotes *A. texanum* from Arkansas. Population IDs by county are as follows: Bedford County (B6), Davidson County (D3), Rutherford County (R1, R7, and R9), Sumner County (S2, S5, S7, and S8), Wilson County (W1, W3, and W4), and Williamson County (Wil2).

Population ID (Map ID)	Sample Size	Latitude	Longituda
Population ID (Map ID)	mtDNA/GBS	Latitude	Longitude
A. barbouri - Tennessee			
Sumner 2 (S2)	3/	36.336487	-86.502042
Sumner 5 (S5)	3/11	36.329443	-86.570217
Sumner 7 (S7)	5/37	36.363339	-86.537930
Sumner 8 (S8)	6/22	36.367510	-86.591100
Davidson 3 (D3)	3/29	36.100821	-86.535713
Wilson 1 (W1)	5/15	36.040205	-86.321052
Wilson 3 (W3)	4/12	36.093415	-86.372530
Wilson 4 (W4)	3/6	36.093249	-86.398001
Williamson 2 (Wil2)	3/5	35.950021	-86.648606
Rutherford 1 (R1)	3/20	35.719718	-86.353396
Rutherford 7 (R7)	12/28	35.672435	-86.315247
Rutherford 9 (R9)	3/11	35.752811	-86.306184
	4/8	35.669700	-86.528800
<u>A. barbouri -</u>			
Kentucky	2/2	27 772270	94 560400
Kentucky FC (FC)	3/3	37.772370	-84.569400
Kentucky (RR)	5/4	37.897710	-84.393000
5 ()	4/3	38.490100	-85.359400
Kentucky SW (SW)	4/5	38.251190	-84.753400
Kentucky JS (JS)	2/2	37.960310	-84.818100
Outgroups			
	2/	33.936860	-78.567865
A. maculatum (AAFB)	2/	35.392500	-86.099722
A. talpoideum (AAFB)	1/1	35.392500	-86.099722
A. texanum (AAFB)	/4	35.392500	-86.099722
A. texanum (AKT1)	1/1	35.828812 -	-90.688313

724

- 726 Table 2. Primers used for PCR amplification of the mitochondrial D-loop. Primers THR (forward) and 651 (reverse)
- 727 were used for initial amplification of entire ~1300 bp. Internal primers 007 (forward), DL3 (reverse), and DL1
- 728 (reverse) were used for Sanger sequencing (Shaffer & McKnight, 1996).

729

Primer Name	Sequence 5'→ 3'
THR (forward)	AACATCGATCTTGTAAGTC
007 (forward)	GCACCCAAAGCAAAATTCTTG
DL3 (reverse)	TTCGATCCAATTGATGAATG
DL1 (reverse)	AATATTGATAATTCAAGCTCCG
651 (reverse)	GTAAGATTAGGACCAAATCT

730

Table 3. Standard measures of genetic diversity for 12 populations of *Ambystoma barbouri* in Tennessee based 586 SNP loci. Summary statistics include the average number of individuals genotyped per locus (N) and the proportion of SNPs that were polymorphic within each population (P). Allelic richness (A_R), observed heterozygosity (H_o), and expected heterozygosity (H_e) were calculated for all loci and again for only those loci that were polymorphic within each

735 population.

36									
				All Loci			Polymor	phic Loci	
37	Population	Ν	Р	A _R	Ho	H _e	A _R	Ho	He
	Sumner 5	9.58	0.108	1.042	0.018	0.020	1.460	0.170	0.154
8	Sumner 7	22.78	0.267	1.036	0.027	0.025	1.204	0.102	0.092
9	Sumner 8	12.97	0.185	1.020	0.025	0.025	1.259	0.136	0.127
55	Davidson 3	22.02	0.192	1.041	0.018	0.016	1.243	0.094	0.083
10	Wilson 1	7.49	0.112	0.986	0.019	0.023	1.395	0.173	0.163
	Wilson 3	7.07	0.045	1.032	0.017	0.016	1.514	0.217	0.184
1	Wilson 4	6.79	0.089	1.049	0.019	0.020	1.602	0.210	0.220
2	Williamson 2	4.11	0.045	1.000	0.012	0.026	1.690	0.257	0.229
	Rutherford 1	16.71	0.092	1.026	0.009	0.008	1.303	0.096	0.091
13	Rutherford 7	23.57	0.140	1.032	0.011	0.011	1.240	0.076	0.079
14	Rutherford 9	7.74	0.056	1.022	0.009	0.009	1.470	0.160	0.159
	Bedford 6	8.74	0.060	1.026	0.010	0.009	1.467	0.167	0.142
15	Average Total	12.46	0.116	1.026	0.016	0.017	1.404	0.155	0.144

750 Table 4. Estimates of effective population sizes (N_E) for five populations of Ambystoma barbouri in Tennessee

based on the Linkage Disequilibrium (LD) method and the Heterozygosity Excess Method as performed by

751 752 N_eEstimator. The 95% confidence intervals were estimated by the parametric chi-squared method.

		Linkage	e Disequilibriu	ım	Heteroz		
Population	N	N_E	CI Lower	CI Upper	N_E	CI Lower	CI Upper
Sumner 7	35	84.5	51.5	207.2	406.4	20.7	∞
Sumner 8	21	107.7	28.8	x	192.7	18.8	∞
Davidson 3	29	52.1	33.2	106.4	24.0	13.0	182.8
Rutherford 1	20	14.8	8.7	30.1	∞	14.6	∞
Rutherford 7	28	30.7	20.9	51.6	∞	22.8	x

753

Table 5. Population pairwise F_{ST} estimates averaged across 584 SNP loci for 12 populations of *Ambystoma barbouri* in Tennessee (below diagonal) and significance of F_{ST} estimates (p-values, below diagonal) estimated from nonparametric permutations of SNP genotypes (100 permutations) as performed by the software Arlequin. ** indicates p < 0.001.

	Populations	1	2	3	4	5	6	7	8	9	10	11	12
1.	Sumner 5												
2.	Sumner 7	0.053											
3.	Sumner 8	0.022	0.055										
4.	Davidson 3	0.650	0.614	0.647									
5.	Wilson 1	0.670	0.616	0.641	0.207								
6.	Wilson 3	0.660	0.590	0.622	0.055	0.104							
7.	Wilson 4	0.581	0.527	0.558	0.111	0.157	0.092						
8.	Williamson 2	0.669	0.568	0.602	0.203	0.238	0.278	0.167					
9.	Rutherford 1	0.725	0.638	0.698	0.288	0.408	0.399	0.320	0.031				
10.	Rutherford 7	0.704	0.637	0.687	0.276	0.234	0.322	0.239	0.077	0.127			
11.	Rutherford 9	0.695	0.601	0.651	0.285	0.356	0.370	0.276	0.039	0.021	0.113		
12.	Bedford 6	0.698	0.600	0.655	0.280	0.386	0.385	0.292	0.000	0.051	0.157	0.085	

Table 6. Results of hierarchical analyses of molecular variation (AMOVA) for the SNP dataset from 12 populations of *A. barbouri* in Tennessee. Assignment to mitochondrial clades are as follows: Clade II (S5, S7, and S8), Clade IV (D3, W1, W3, and W4), and Clade III (Wil2, R1, R7, R9, B6). Asterisks indicate significance of Φ statistics based on 1000 permutations in Arlequin.

Source of variation	DF	Sum of squares	Variance components	Percentage of variation	Φ-Statistics
Among drainages	2	614.5	2.30	63.60	$\phi_{CT} = 0.636*$
Among populations within drainages	9	31.24	0.08	2.09	$\phi_{SC} = 0.058*$
Within populations	384	476.24	1.24	44.23	$\phi_{ST} = 0.657*$
Total	395	1,121.97	3.62		

Figure 2. Map of all surveyed populations of *A. barbouri* in Tennessee and Kentucky. County lines are shown with black borders. Shaded regions indicate major watersheds as depicted in the legend. Tennessee populations are colored by ESU assignment where green circles indicate northern ESU populations, yellow circles indicate central ESU populations, and purple circles indicate southern ESU populations.

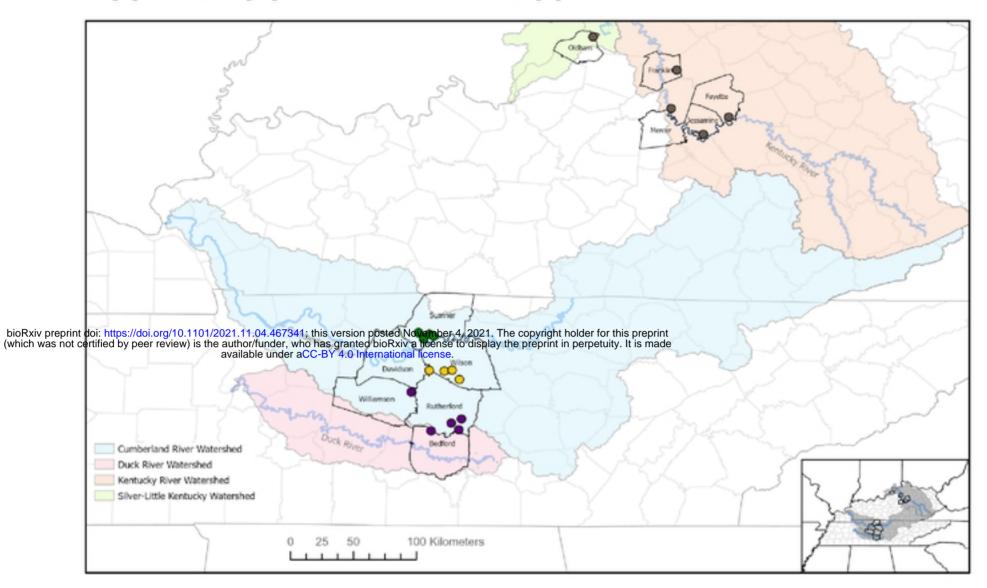
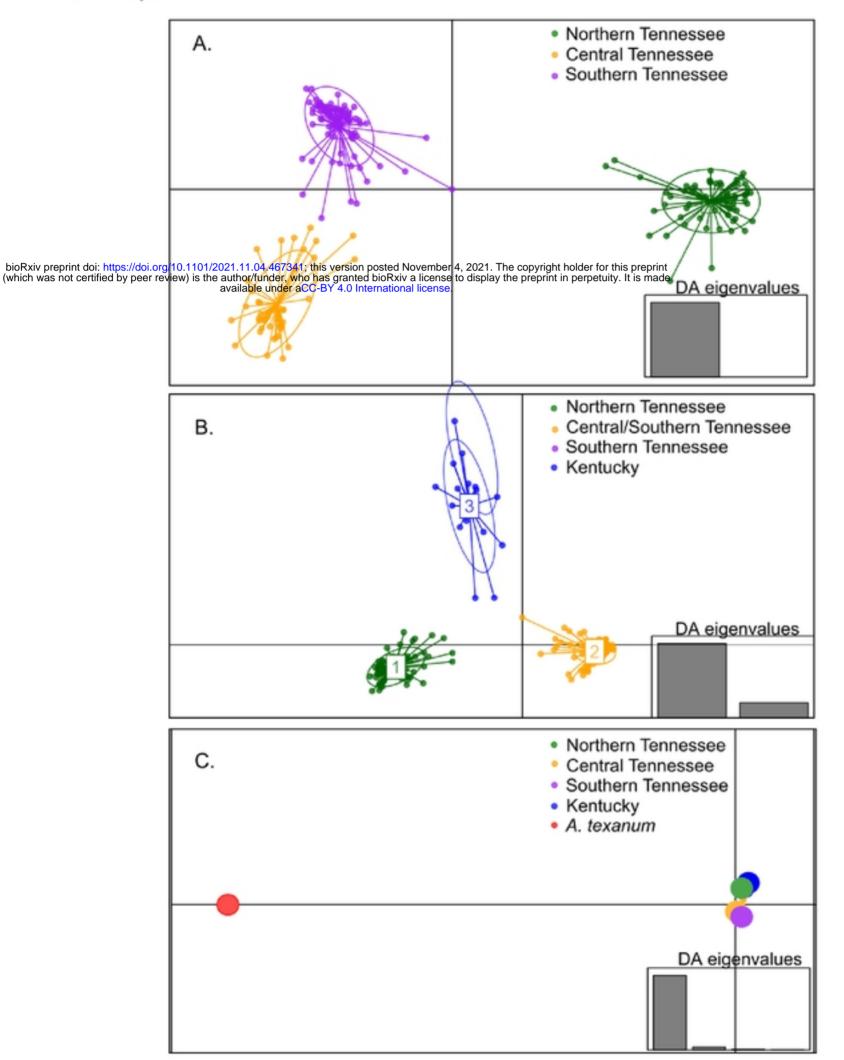
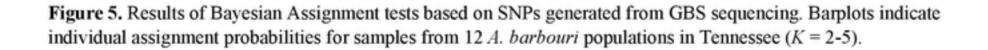




Figure 4. Discriminant Analysis of Principal Components (DAPC) results based on SNP genotypes A.) DAPC plot for *A. barbouri* (N = 198) from Tennessee. Analysis assumed a value of K = 3 as the optimized by BIC analysis. B.) DAPC plot for *A. barbouri* from Tennessee (N = 198) and also including samples from five populations of *A. barbouri* in Kentucky (N = 17). Analysis was performed using K = 3 as the optimal number of genetic clusters based on BIC analysis. C) DAPC plot for *A. barbouri* from Tennessee, Kentucky, and also two populations of *A. texanum* (N = 4). Analysis was performed using K = 5 as the optimal number of genetic clusters according to the BIC analysis.



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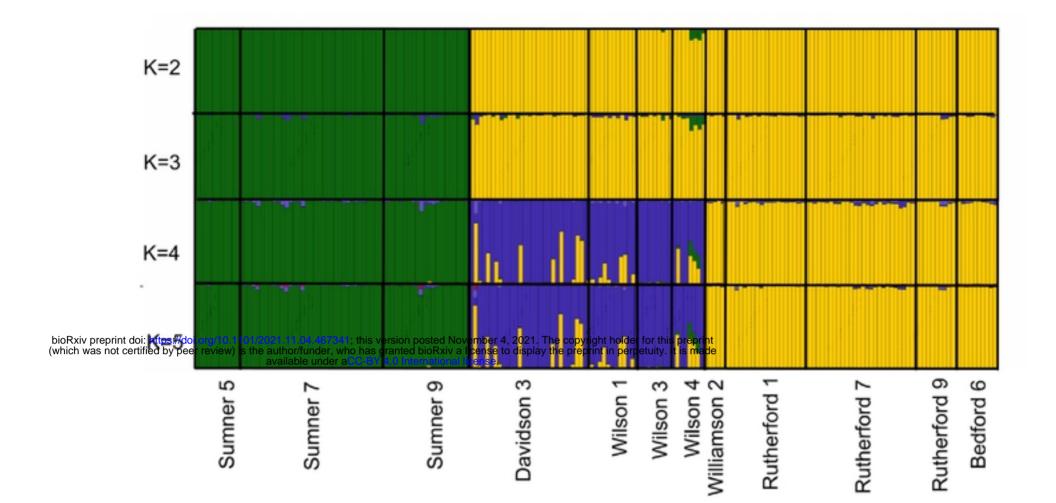
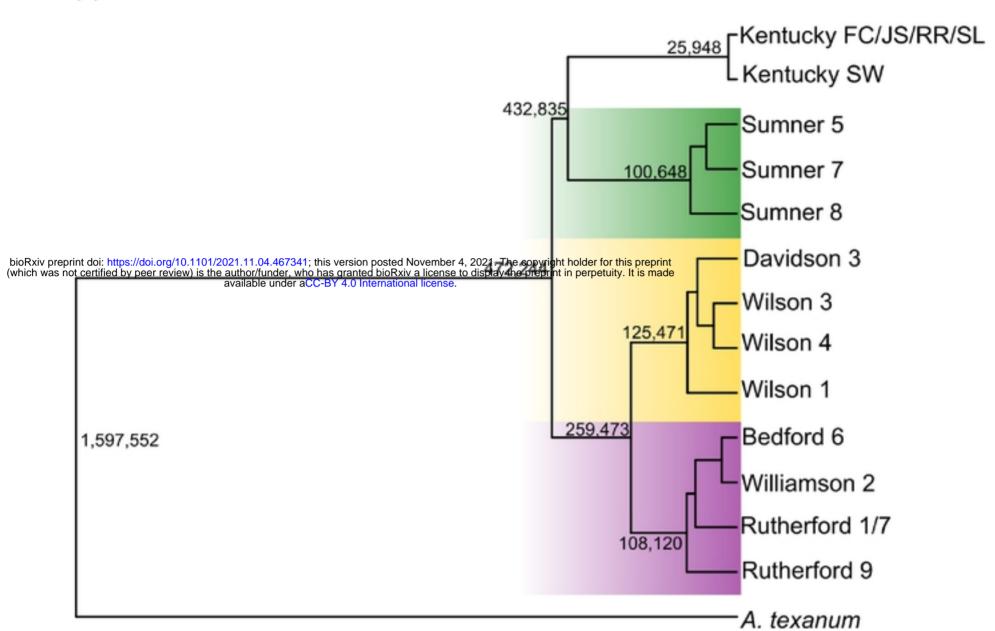


Figure 6. Species-tree (A) and divergence time estimates (B) from Bayesian multispecies coalescent method as implemented in the software BPP. Numbers at nodes also indicate divergence time estimates (*t*). Divergence time parameter τ was converted to time in years before present (YBP) in the BPPR statistical package in R-studio. Shading in tree indicates ESU assignments as follows: Northern ESU (green), Central ESU (Yellow) and Southern ESU (Purple).

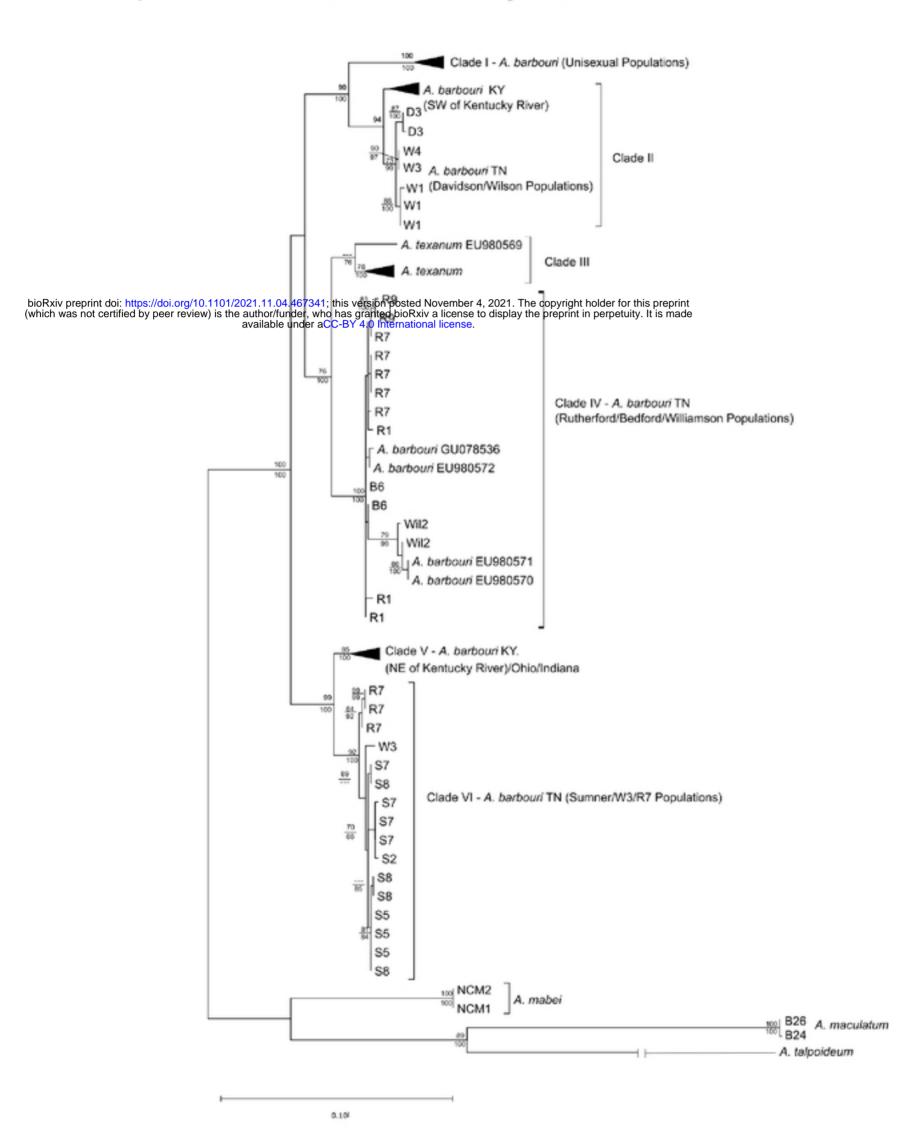


(A)

(B)

Node	TMRCA (YBP)	CI (95%)		
A. texanum, A. barbouri	1,597,552	1,169,671	2,062,859	
A. barbouri (all)	472,244	313,795	639,208	
Kentucky (all), Sumner (all)	472,244	313,795	639,208	
Kentucky (all)	24,982	13,432	41,552	
Sumner (all) Davidson, Wilson, Williamson, Bedford,	100,648	45,292	162,312	
Rutherford	259,743	175,719	350,108	
Davidson, Wilson	125,471	75,781	195,323	
Bedford, Williamson, Rutherford	108,120	62,980	164,210	

Figure 3. Maximum likelihood phylogenetic reconstructions of unique mitochondrial D-loop haplotypes from *Ambystoma barbouri* and *A. texanum* under a GTRCAT model of evolution as performed by RaxML. Bootstrap support values above branches are shown for nodes with 70% support or greater. Values below nodes indicate posterior probabilities from Bayesian reconstructions under a T92+G model of sequence evolution as performed by MrBayes. Asterisks denote accession numbers for sequences downloaded from Genbank.





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Figure 1. Photo image of Tennessee Ambystoma barbouri.

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Figure 1 - barb pic