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2 Phylogeny and species delimitation of Near Eastern <i>Neurergus</i> newts (Salamandridae)
3 based on genome-wide RADseq data analysis
4
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23Short title: Phylogeny of <i>Neurergus</i> newts
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35 36**Abstract**

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38We reconstruct the molecular phylogeny of Near Eastern mountain brook newts of the genus 39Neurergus (family Salamandridae) based on newly determined RADseq data, and compare 40the outcomes of concatenation-based phylogenetic reconstruction with species-tree inference. 41Furthermore, we test the current taxonomy of *Neurergus* (with four species: *Neurergus* 42strauchii, N. crocatus, N. kaiseri, and N. derjugini) against coalescent-based species-43delimitation approaches of our genome-wide genetic data set. While the position of *N*. 44strauchii as sister species to all other *Neurergus* species was consistent in all of our analyses, 45the phylogenetic relationships between the three remaining species changed depending on the 46applied method. The concatenation approach, as well as quartet-based species-tree inference, 47supported a topology with N. kaiseri as the closest relative to N. derjugini, while full-48coalescent species-tree inference approaches supported *N. crocatus* as sister species of *N*. 49*derjugini*. Investigating the individual signal of gene trees highlighted an extensive variation 50among gene histories, most likely resulting from incomplete lineage sorting. Coalescent-51based species-delimitation models suggest that the current taxonomy might underestimate the 52species richness within *Neurergus* and supports seven species. Based on the current sampling, 53our analysis suggests that *N. strauchii*, *N. derjugini* and *N. kaiseri* might each be subdivided 54into further species. However, as amphibian species are known to be composed of deep 55conspecific lineages that do not always warrant species status, these results need to be 56cautiously interpreted in an integrative taxonomic framework. We hypothesize that the rather 57shallow divergences detected within *N. kaiseri* and *N. derjugini* likely reflect an ongoing 58speciation process and thus require further investigation. On the contrary, the much deeper 59genetic divergence found between the two morphologically and geographically differentiated 60subspecies of *N. strauchii* leads us to propose that *N. s. barani* should be considered a distinct 61species, N. barani (Öz 1994).

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64*Keywords: Neurergus*, Multi-Species Coalescent, Species-tree inference, Species 65delimitation, ddRAD sequencing, *Neurergus barani*. 66

681. Introduction

69Mountain brook newts of the genus *Neurergus* (Cope 1862) form a monophyletic group 70occurring in the Near East, from Turkey to Iraq and Iran. Although a few studies have 71 focused on different aspects of *Neurergus* biology (Bogaerts et al., 2006; Hendrix et al., 2014; 72Schneider and Schneider, 2010; Sparreboom et al., 2000; Steinfartz et al., 2002), our 73knowledge on their ecology, behavior and evolutionary history remains – compared to other 74salamandrid genera such as European newts and salamanders – understudied, mainly because 75of their remote ranges in countries difficult to access (Hendrix et al., 2014). All Neurergus 76species inhabit geographically restricted areas (Fig. 1) and are included in threatened 77 conservation categories, ranging from vulnerable to critically endangered (IUCN SSC 78Amphibian Specialist Group, 2016; Papenfuss et al., 2009; Papenfuss et al., 2009; Sharifi et 79al., 2009). Thus, it is important to elucidate their biology and ecology in order to inform 80effective conservation strategies. In this context, knowledge on the phylogeny, systematics 81and taxonomy of these threatened amphibians is key to delimiting where conservation efforts 82need to focus. Fine-scale phylogenetic and phylogeographic studies can be used to delimit 83such conservation units (Vogler and Desalle, 1994), and furthermore provide the means to 84analyse the genetic structure of populations as a basis for downstream conservation measures 85(e.g. Dincă et al., 2018). As conservation assessments and legal documents traditionally use 86species as principle conservation units, it is also important to have a correct species-level 87taxonomy, particularly when cryptic, or near-cryptic, species are involved (Delić et al., 2017).

In *Neurergus*, four species are currently recognised based on molecular 89(mitochondrial and nuclear genes) and morphological characters: *N. crocatus*, *N. derjugini* 90(including the subspecies *derjugini* and *microspilotus*), *N. kaiseri*, and *N. strauchii* (including

91the subspecies *barani, strauchii* and *munzurensis*). The monophyly of the genus has been 92well established based on nuclear allozyme and mitochondrial DNA sequence data (Steinfartz 93et al., 2002) and phylogenetic reconstructions within the Salamandridae, based on complete 94mitogenomes and nuclear sequences (Kieren et al., 2018; Veith et al., 2018; Zhang et al., 952008), placed it as sister group to Near Eastern newts of the genus *Ommatotriton*. The 96relationships within *Neurergus* have also been studied using a variety of molecular markers 97(Hendrix et al., 2014; Steinfartz et al., 2002), but always based on a limited number of loci. 98Consequently, evolutionary relationships among some *Neurergus* species remain poorly 99resolved and the taxonomic status of the *N. derjugini* and *N. strauchii* subspecies *N. derjugini* 100and *N. strauchii* remain uncertain.

Considering the enormous methodological progress in the genomic era, it has 102become possible to resolve phylogenetic relationships and to investigate species boundaries 103based on markers representing evolutionary diversification processes at different parts of an 104organism's genome. A variety of approaches exist to sample loci at the genomic scale for 105such purposes. Among other methods, it is possible to sequence only the expressed regions of 106the genome (RNA sequencing or RNAseq, Ozsolak and Milos, 2011), to target specific 107regions of the genome (e.g. exon capture, Bragg et al., 2016; enrichment of Ultra Conserved 108Elements or UCEs, Faircloth et al., 2012), or to sequence randomly small DNA fragments 109(e.g. Restriction site Associated DNA sequencing, Miller et al., 2007; Genotyping By 110Sequencing, Elshire et al., 2011). However, it can be challenging to ensure homology of loci 111across taxa and to verify assumptions such as absence of selection and of intra-locus 112recombination. Thus, a careful choice of markers is paramount, particularly for poorly known 113non-model organisms.

114 Restriction site Associated DNA sequencing (RADseq) has become popular in 115recent years for phylogenetic, population genetic and systematic studies, as it can generate 116sequences of a large number of independent loci for a moderate cost (Andrews et al., 2016). 117In brief, genomes are digested with one or several restriction enzymes, and a short DNA 118 fragment flanking the restriction site is sequenced, resulting in hundreds of small fragments 119that make a reduced representation of the whole genome available to analysis (Miller et al., 1202007). Even though this method was originally used for inferences at the intraspecific level 121(Ree and Hipp, 2015), both empirical (e.g. Ebel et al., 2015; Manthey et al., 2016; Razkin et 122al., 2016) and in silico (e.g. Cariou et al., 2013; Rubin et al., 2012) studies have shown its 123usefulness to reconstruct well resolved phylogenies, even at relatively deep evolutionary 124scales (e.g. Lecaudey et al., 2018), with performances comparable to other genomic 125approaches (e.g. full genomes: Cariou et al., 2013; UCE targeting: Manthey et al., 2016; 126RNAseq: Rodríguez et al., 2017). As in more classical multilocus approaches, RADseq loci 127can be concatenated to infer phylogenies. The resulting concatenation matrices are usually 128very large (hundred of thousands to millions of bp), holding numerous informative sites, and 129thus generally lead to accurate inferences (Ree and Hipp, 2015). However, in cases of 130variation among gene histories, species-tree inference may become necessary for correct 131phylogenetic reconstruction, and different approaches are available for this purpose.

132 The more popular methods (e.g. *BEAST, Heled and Drummond, 2010; ASTRAL, 133Mirarab et al., 2014) use gene trees (which are jointly estimated or taken as input) to infer a 134species-tree that is consistent under the multi-species coalescent model. However, gene trees 135inferred from RADseq loci usually have very low resolution as the sequences are very short 136(Harvey et al., 2016) and the high number of loci makes the analyses very long and intensive. 137Methods inferring species-trees from bi-allelic markers, without explicitly estimating gene

138trees (e.g. SNAPP, Bryant et al., 2012; SVDquartets, Chifman and Kubatko, 2014) could be 139more appropriate to RADseq data as these can sample unlinked SNPs at a genomic scale 140(Peterson et al., 2012). Yet, it is still unclear which method should be best applied for 141RADseq data at a given evolutionary scale. This uncertainty also applies to coalescent-based 142species-delimitation, as one of the most popular approaches, BPP (Yang, 2015), requires a 143resolved species-tree, either estimated from gene trees or specified by the user. This method 144has been used with RADseq data (e.g. Nieto-Montes de Oca et al., 2017), but its ability to 145reconstruct the true species-tree from these data, as well as the impact of errors in the species-146tree on the species-delimitation, are not clear. Here, we apply a RADseq approach to sample 147several thousand loci across the genome of all described species of *Neurergus*. Using these 148data, we (i) infer the phylogeny of the genus from a matrix of concatenated loci and from (ii) 149various alternative species-tree methods, and (iii) investigate species boundaries using several 150coalescent-based species-delimitation approaches. The aim of this study is to resolve the 151phylogeny and systematics of this understudied genus, but also to explore new empirical 152insights on the performance and congruence of alternative coalescent-based species-tree 153reconstruction and species-delimitation methods on RADseq data.

154

1552. Material and Methods

1562.1 Sampling and laboratory work

157Tissue samples were collected from all the species and subspecies of *Neurergus* (except *N*. *s*. 158*munzurensis*), for a total of 17 individuals from 13 locations (Fig. 1). Due to different 159sampling efforts, the distribution of samples is unequal among species, and especially 160samples of *N. kaiseri* are overrepresented. Additionally, one individual of both *Triturus* 161*cristatus* and *Ommatotrion vittatus* were sampled as outgroups. Genomic DNA was extracted

162using the Macherey-Nagel NucleoSpin Tissue kit following the manufacturer's instructions. 163We performed double-digest Restriction Site Associated DNA sequencing (ddRADseq; 164Peterson et al., 2012) preparing the library as follows (per Recknagel et al., 2015 with 165modification of Illumina adapters): 1 μg of DNA from each individual was double-digested 166using the PstI-HF and AcII restriction enzymes (NewEngland Biolabs); modified Illumina 167adaptors with unique barcodes for each individual were ligated onto this fragmented DNA; 168samples were multiplexed (pooled); and a PippinPrep was used to size select fragments 169around a tight range of 383 bp, based on the fragment-length distribution identified using a 1702200 TapeStation instrument (Agilent Technologies). Finally, enrichment PCR was 171performed to amplify the library using forward and reverse RAD primers. Sequencing was 172conducted on an Illumina Next-Seq machine at Glasgow Polyomics to generate paired-end 173reads 75 bp in length.

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1752.2 Assembly of RADseq data

176De-multiplexed reads were assembled *de novo* using ipyrad (Eaton, 2014; Eaton, 2015). This 177method clusters reads depending on their similarity and then calls bases above a user-defined 178minimum depth. In order to avoid introducing errors in the dataset or clustering paralogous 179loci, we carefully investigated the impact of these parameters on the number of recovered loci 180and the phylogenetic accuracy of the obtained dataset. First, we assembled the dataset with a 181minimum depth ranging from 2 to 10, with the clustering threshold set as 85%. Then, once the 182minimum depth was selected, we performed 10 assemblies with clustering thresholds ranging 183from 85% to 95%. For the values of 85%, 90% and 95%, we filtered the datasets with 184different minimum numbers of individuals per locus (4, 8, 10, 14, 16 and 18, corresponding to 18520%-95% of individual coverage) and inferred phylogenies using RAxML (Stamatakis, 2014) 186with 100 rapid bootstraps under a GTR+GAMMA substitution model on the concatenation of 187all loci, and the concatenation of unlinked SNPs (i.e. one SNP sampled at each locus, 188corresponding either to the one with the highest coverage or, in case of equal coverage, 189randomly chosen). As some analyses do not require the inclusion of outgroups (e.g. some 190species-tree inference methods), we also assembled a dataset containing only the *Neurergus* 191samples. Indeed, removing distant species should increase the amount of data recovered and 192the accuracy of the analyses (Pante et al. 2015). For this second dataset, we performed the 193same series of tests than for the first one, with minimum individual coverage ranging from 4 194to 16 (23%-94% of the total number of individuals). In the end, a total of 66 different datasets 195were assembled and trees inferred (Tab. S1). For each of the trees, the average bootstrap 196support of the "deep nodes" (i.e. relationship between the Neurergus species) and the 197" shallow nodes" (i.e. nodes at specific and intra-specific level) were calculated. The 198combination of parameters maximizing the bootstrap support of the phylogeny and the 199amount of loci recovered was then used for the final analyses. After performing the final 200assemblies of both datasets, ipyrad was used to format the data in different ways: 201 concatenation of all the loci, sequences for each locus, and concatenation of unlinked SNPs 202extracted from the loci.

203

2042.3 Concatenation analysis

205The best model of substitution for the concatenated loci was selected using the Bayesian 206Information Criterion (BIC) implemented in ModelGenerator (Keane et al., 2006). 207Phylogenetic relationships were then inferred using RAxML with the new rapid hill-climbing 208algorithm on the concatenated matrix, with *Triturus cristatus* and *Ommatotriton vittatus* as 209outgroups. 100 bootstrap replicates were performed to assess the support of the nodes.

210Because for extensive molecular datasets, bootstrap values can sometimes be misleading and 211provide high support for spurious nodes (Rodríguez et al., 2017; Rubin et al., 2012), we also 212performed gene jackknifing (Delsuc et al., 2008): loci were randomly sampled up to ca. 21310,000, 100,000, 500,000 and 800,000 positions, and 100 random draws were made for each 214value. For each replicate, RAxML was run with the same setup as previously, and the 215proportion of replicates supporting each node of the complete tree was calculated using the 216bipartition model implemented in RAxML. In order to quantify the genetic divergence 217between the phylogenetic lineages, pairwise nucleotide diversities between all groups were 218calculated as the average across all loci using the R package PopGenome (Pfeifer et al., 2192014).

220

2212.3.2 Species tree inference

222We inferred the species-tree using various approaches. First, we used SNAPP (Bryant et al., 2232012), a method to infer a species-tree from unlinked SNPs without estimating gene trees. 224This model uses a Bayesian approach that is implemented in BEAST2 (Bouckaert et al., 2252014). For this analysis, individuals were clustered into putative species on the basis of the 226concatenation tree. As this analysis implements the coalescent model which does not require 227rooting to polarize the topology, it was run on the dataset without outgroups, with 15,000,000 228mcmc steps and sampling every 10,000 steps, and with a G (2, 1000) prior distribution on 229population size and default priors on mutation rates (mutation rate U = 1.0, mutation rate V = 2301.0). The convergence of the analysis was checked using Tracer 1.7 (Rambaut et al., 2018). 231We then implemented two methods that reconstruct the species-tree from quartet trees. Tetrad 232(Eaton, 2015b) features a method similar to SVDquartets (Chifman and Kubatko, 2014): 233quartet trees are inferred from unlinked SNPs, and joined together into a species-tree that is 234consistent under the multi-species coalescent. Thus, the analysis was first performed on the 235dataset including the two outgroups, with all possible quartets sampled, and with 500 non-236parametric bootstrap replicates. To test for the effect of the amount of data on the support of 237the nodes, the method was also performed on the dataset without outgroups, with the same 238parameters. We also used ASTRAL-II (Mirarab and Warnow, 2015), which implements a 239similar method, but the quartet trees are calculated from gene trees. Thus, the gene trees for 240all loci were inferred from the *Neurergus*-specific dataset, using RAxML with the rapid hill-241climbing algorithm and a JC69 substitution model, and with 100 bootstrap replicates for each 242gene tree. Then, the species-tree was inferred using ASTRAL-II, and again 100 bootstrap 243replicates were performed.

244

2452.3.3 Phylogenetic signal of gene trees

246To resolve controversial relationships, the signal of individual gene trees was investigated. 247The two conflicting topologies differ in the position of *N. kaiseri* as the sister taxon of *N.* 248*derjugini* and *N. d. microspilotus* (Concatenated, tetrad and ASTRAL-II; see Results), or as 249*N. crocatus* as the sister taxon to *N. derjugini* and *N. d. microspilotus* (BPP and SNAPP; see 250Results). Under ILS alone, one of the topologies described above should represent the true 251history and occur in higher frequency in the population of gene trees. Thus, the proportion of 252genes supporting each alternative topology, as well as a third alternative with *N. crocatus* as 253sister species to *N. kaiseri*, was calculated with a custom R script, using the package ape 254(Paradis et al., 2004). We used the same gene trees as for the ASTRAL-II analysis. For each 255gene tree we sampled one individual per species and verified the monophyly of each 256alternative group. We ran 100 independent replicates of this analysis to capture the variation 257generated by sampling only one individual per species. We used ASTRAL-II to compute the quartet scores of the different competing 259topologies. This is another way to measure the overall agreement of the gene trees with a 260topology; the quartet score is the proportion of quartet trees induced by all the gene trees that 261are present in the species-tree (Mirarab et al., 2014).

262

2632.4 Species delimitation

264The putative species used in the SNAPP analysis were tested using the software BPP 265(Bayesian Phylogeny and Phylogeography; Yang, 2015; Yang and Rannala, 2010). The 266species-tree and species-delimitation were jointly inferred (analyze A11). A gamma 267distribution G(2, 1000) was used as a prior on the population-size parameter, and the 268distributions G(2, 100) and G(2, 10000) were also tested to check for the effect of the prior 269on the results. A gamma distribution G(10, 1000) was used as a prior on the age of the root of 270the tree, corresponding to the time of divergence of the *Neurergus* inferred from 271mitochondrial markers by the study of Steinfartz et al. (2002), and the other splitting times 272were inferred using a dirichlet process. The analysis was set with 100,000 burnin steps and 273200,000 steps, with sampling every 10 steps. In case the species-tree inferred with BPP 274differed from the one inferred with tetrad, the latter was used to perform species-delimitation 275with a fixed guide tree with BPP (analysis A10) to test for the effect of uncertainty on the 276species-tree on the species-delimitation. Due to limited computational resources, the whole 277dataset could not be analyzed at once. Thus, 5 datasets of 500 randomly drawn loci were built 278and analyzed independently, a method first used by Nieto-Montes de Oca et al. (2017).

279 Species boundaries were also investigated using BFD* (Leaché et al., 2014). This 280method computes the marginal likelihood of species-trees inferred with SNAPP using the 281PathSampling algorithm implemented in BEAST2. Different models can be tested by

282lumping and splitting populations or reassigning individuals, and their marginal likelihood 283can be compared to find the best fitting model. Alternative species delimitation hypotheses 284were defined using the concatenation tree (Table 3), and their marginal likelihood was 285calculated by sampling 48 steps with the following parameters: 400,000 MCMC steps with 28650% burnin and 10,000 pre-burnin steps. MCMC parameters for species-tree inference were 287left as for the previous SNAPP analysis. For each model, two independent runs were 288performed, and the average Marginal Likelihood was retained.

289

2903. Results

2913.1 RADseq dataset reconstruction

292Processed reads ranged from 781,375 to 9,365,666 per sample. The characteristics of the 66 293datasets assembled are given in Table S1. Varying the clustering threshold from 85% to 95% 294had mostly no impact on the topology inferred. Out of the 66 trees inferred, 63 of them 295yielded the topology given in Fig. 2, usually with high bootstrap supports, with the three 296inconsistent topologies resulting from assemblies with very few loci (unlinked SNPs with 297high individual coverage, Fig. S1). Among these three, two placed *N. crocatus* as sister 298species to *N. kaiseri*, and one failed to properly recover the intra-specific structure of *N*. 299*kaiseri*. The final dataset was assembled using a minimum depth of 8, a clustering threshold 300of 89%, and a filtering threshold that excluded all loci with fewer than 16 individuals (14 for 301the dataset without outgroups, respectively corresponding to 82% and 84% of the 302individuals). Finally, we recovered 8,314 loci with an average length of 115 bp (min: 109 bp; 303max 127 bp). The final matrix includes 958,256 bp, with a total of 12.24% missing data (see 304Table 1 for more details). As expected, removing the outgroup taxa significantly increased the

305number of loci: the *Neurergus*-specific dataset contained 15,547 loci (1,791,783 bp). From 306these two datasets, respectively, 8,314 and 14,765 supposedly unlinked SNPs were extracted. 307

Dataset	Length of the concatenation	Number of loci	Number of unlinked SNP	Percentage of missing data	Number of variable sites	Number of parsimony informative sites
With outgroups	958,256 bp	8,314	8,314	12.24%	40,781 (4.3%)	25,799 (2.7%)
Without outgroups	1,791,783 bp	15,547	14,765	9.3%	52,795 (2.9%)	44,343 (2.5%)

308Table 1. Summary	statistics	for the	two datasets	assembled.
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3113.2 Phylogenetic analyses

3123.2.1 Concatenation analyses

The phylogenetic tree inferred from the concatenated matrix of all loci was fully 314resolved, with 100% bootstrap support at every node (Fig. 2). Gene jackknifing also 315recovered full support for all nodes from ca. 100,000 bp sampled. The tree can be roughly 316divided in two groups: *N. strauchii*, with a deep divergence between the subspecies *strauchii* 317and *barani* (average pairwise nucleotide diversity = 0.0056), and a monophyletic group 318including *N. crocatus*, *N. derjugini* and *N. kaiseri*. Within this clade, *N. derjugini* and *N*. 319*kaiseri* are sister lineages. The subspecies of *N. derjugini* show very shallow divergence 320(smaller than the intraspecific divergence between the two localities of *N. crocatus*; average 321pairwise nucleotide diversity = 0.0035), but are reciprocally monophyletic with high support. 322*N. kaiseri* displays a high genetic diversity between localities and can be split into two main 323clades, corresponding to locations found in the southern (Ab Palangi and Shevi fall) and 324northern (Bozorg Ab, Barik Ab, Tova, Shahbazan and Nergeseh) parts of the distribution 325range. For the subsequent species-tree and species-delimitation inferences, all described326species and subspecies, as well as both clades of *N. kaiseri*, were considered putative species.327The matrix of the pairwise nucleotide diversities is displayed in Table 2.

328

	N. s. strauchii	N. s. barani	N. crocatus	N. d. derjugini	N. d. microspilotu s	N. kaiseri N
N. s. strauchii						
N. s. barani	0.0056					
N. crocatus	0.0073	0.0074				
N. d. derjugini	0.0075	0.0077	0.0071			
N. d. microspilotu s	0.0076	0.0077	0.0071	0.0035		
N. kaiseri N	0.0072	0.0073	0.0066	0.0067	0.0068	
N. kaiseri S	0.0074	0.0076	0.0070	0.0072	0.0073	0.0039

329Table 2.	Nucleotide	diversitv	between all	pairs of lineages	s, averaged on	all the loci
0-0 -0.010 -0	1 . acreotade		occir cen un	pano or micage	, a, cragea on	and the root

330 331

3323.2.2 Species-tree inference

333The topology of the species-tree inferred with tetrad was consistent with the concatenation
334tree (hereafter referred to as "topology A"; Fig. 3a). All the nodes received full support,
335except for the one at the root of the group including *N. crocatus*, *N. derjugini* and *N. kaiseri*,
336supported by only 74% of the bootstrap replicates. When running the analysis on the dataset
337without outgroups, the topology remained the same and the support of this node increased to

33894%. ASTRAL-II revealed the same topology, except that *N. d. derjugini* was paraphyletic. 339The tree inferred with SNAPP, on the other hand, showed a different topology (hereafter 340referred to as "topology B"; Fig. 3b), placing *N. kaiseri* as sister species of a clade constituted 341of *N. crocatus* and *N. derjugini*. All the nodes had a posterior probability of 1. However, even 342with 15,000,000 steps (running for a total of ca. 154 h using 10 cores), the MCMC run did 343not converge properly as the trace of the topology posterior probability showed a clearly 344increasing trend, and the ESS remained very low. When using BPP to infer the species-tree 345jointly with species-delimitation, we also obtained "topology B", and with good convergence 346of the MCMC chain.

347

3483.2.3 Phylogenetic signal of gene trees

349To understand the support from single markers for each of the alternative topologies A and B, 350the signal of the gene trees was investigated. An average of 18% of the loci supported 351topology B, while 17% supported topology A (Fig. 4). Additionally, 13 % of loci supported 352the position of *N. crocatus* as sister species to *N. kaiseri*. Topology A obtained a quartet score 353of 0.601, and topology B of 0.636.

354

3553.3 Species-delimitation

356Using BPP, the seven *a priori* defined taxa were all delimited as different species with high 357posterior probability (i.e. P > 0.95) in 17 analyses out of the 20 performed. In the three other 358cases, the two subspecies of *N. derjugini* were lumped into a single species, sometimes with 359full support (P=1). Variations on the theta prior had no significant impact on the species 360delimitation. The species-tree inferred with BPP was identical to the one inferred with 361SNAPP (topology B). Using the alternative topology as fixed guide tree to perform species-

362delimitation did not change the results.

363 The BFD* analyses supported the BPP results, as the seven-species model obtained 364the highest marginal likelihood (Table 3).

365

366**Table 3.** Summary of species-delimitation models and their respective marginal likelihood 367values, tested using BFD*, ranked by decreasing marginal likelihood.

	Number of species	Model	Marginal L
	7	Split N. strauchii, N. derjugini and N. kaiseri	-94087.385
	6	Split <i>N. strauchii</i> and <i>N. kaiseri</i>	-94616.512
	6	Split <i>N. derjugini</i> and <i>N. kaiseri</i>	- 95608.314
	5	Split N. kaiseri	-96079.767
	6	Split <i>N. strauchii</i> and <i>N. derjugini</i>	-96290.751
	5	Split N. strauchii	-96875.032
	5	Split <i>N. derjugini</i>	-97900.825
	4	Current taxonomy	-98270.815
368			

369 370**4. Discussion**

3714.1 Phylogenetic inference and species-delimitation of Neurergus using RADseq data

372Our first approach to infer phylogenetic relationships among species of *Neurergus* relied on a 373concatenation of all RAD loci, followed by the analysis of this supermatrix with classical 374tools such as RAxML. As expected, the resulting matrix was rather large and informative, and

375its analysis yielded a fully resolved tree with 100% bootstrap support for every node 376(topology A; Fig. 2). Gene-jackknifed pseudo-replicates of the concatenated dataset also 377recovered full support for every node from ca. 100,000 bp sampled. We then used several 378species-tree reconstruction approaches as alternative means to assess whether this tree could 379be considered representative of the evolutionary history of these newt species. The two 380species-tree approaches based on quartet trees, tetrad and ASTRAL-II, resulted in the same 381topology as the concatenation analysis (Fig. 3a). Nodes were overall strongly supported, 382except for the one defining the *N. crocatus* + *N. derjugini* + *N. kaiseri* clade, which in the 383tetrad analysis received quite low support. However, when using the most complete dataset 384(without outgroup taxa, 14,765 SNPs), the support of this node increased substantially, thus 385suggesting that the lower support was due to the lack of informative loci and not to genuine 386variation among gene trees. On the contrary, two other coalescent-based methods for species-387tree inference yielded a different, yet strongly supported topology (Fig. 3b). Indeed, in both 388SNAPP and BPP results, *N. derjugini* comes out as sister species to *N. crocatus*, and not to *N.* 389kaiseri as in the previous analyses.

Discrepancies between concatenation trees and species-trees are well documented 391(Kubatko and Degnan, 2007; Roch and Steel, 2015) and are usually attributed to four main 392processes generating variation among gene genealogies: (i) Incomplete Lineage Sorting 393(Pollard et al., 2006); (ii) gene flow between lineages (Solís-Lemus et al., 2016); (iii) gene 394duplication and loss; and (iv) recombination (Degnan and Rosenberg 2006). In the present 395case, both competing topologies had a rather low quartet score, highlighting a high level of 396discordance among gene trees. Further investigations revealed that topology A was supported 397by a proportion of ca. 17 % of the gene trees, while ca. 18 % supported topology B. Under 398the coalescent, short internal branch lengths in the species-tree can lead to the predominance

399of gene tree topologies that differ from the species-tree history. This phenomenon results in 400high support for the wrong topology if the genes are concatenated, generating a conflict 401between species-tree and concatenation methods. Such cases of rapid diversification in 402internal branches of the species-tree history are defined to be in the gene tree anomaly zone 403(Kubatko and Degnan 2007; Degnan 2013). As the branch at the root of the *N. crocatus* + *N.* 404*kaiseri* + *N. derjugini* group is rather short, suggesting a rapid divergence, we could expect a 405higher proportion of gene trees supporting the wrong species-tree topology. Here, topology B 406was supported by more gene trees than topology A, although the difference was very low. The 407theoretical expectation in the anomaly zone would be to have a high support for topology B if 408the loci are concatenated, which is contrary to our results. Furthermore, the small difference 409in gene tree support for each competing topologies (17% versus 18%) could be a 410consequence of other evolutionary mechanisms, like gene flow among the lineages. Given the 411isolated geographic distribution of the sampled populations, current gene flow is highly 412improbable and can be excluded with some confidence, but migration between ancestral 413populations could have happened. In this case, the tree-like representation of the relationships 414 within the group would be misleading, and the evolutionary history of the Neurergus would 415be best represented as a network with a deep reticulation (Degnan 2018; Burbrink and Gehara 4162018). As the coalescent methods applied here only account for ILS as a source of variation 417among gene trees, it is impossible to rule out other sources of discordance with certainty. 418 On the other hand, the discrepancy between species-tree inference methods is more 419complicated and could be related to a combination of biological processes and differences in

420the mechanics of each method, or intrinsic characteristics of the different datasets used. The 421impact of the assembly and filtering parameters on downstream analyses has been 422investigated in several studies (e.g. Leaché et al. 2015, Lee et al. 2018). A low clustering

423threshold, for example, can have a strong impact on phylogenetic analyses, by clustering 424paralogous or non-homologous loci together (Rubin et al. 2012). Similarly, the amount of 425missing data in RADseq datasets, either due to low individual coverage of the loci, or loci 426dropout when distantly related species are included (Viricel et al., 2014), can be problematic, 427for example for species-tree inference (Schmidt-Lebuhn et al., 2017). However, in our case, 428the topology inferred was very constant when varying the clustering threshold, the minimum 429individual coverage of the loci, and the presence or absence of outgroups (Tab. S1, Fig. S1). 430Thus, the processing of the RADseq data seems not to be the cause of our conflicting results.

431 With respect to the differences in each analytical method, it is important to note that 432the MCMC of the SNAPP analysis did not mix properly and had low ESS values. Also, the 433BPP analysis was performed with fractions of the dataset due to computational limitations. 434These two methods apply the full Bayesian function, modeling the entire evolutionary 435process, from sequences or SNPs to species-trees, taking into account phylogenetic 436uncertainty in the gene tree topologies and parameters of the multispecies coalescent model. 437In theory, they are more robust than methods that do not full Bayesian, like ASTRAL-II and 438tetrad. Both SNAPP and BPP inferred a different topology and are both consistent with each 439other. However, these results should be taken with caution because of poor chain mixing in 440SNAPP and the smaller amount of data used in BPP. Therefore, additional analyses will be 441needed to understand the evolutionary processes within *Neurergus* newts. Particularly, less 442computationally intensive methods that jointly estimate gene trees and species-trees, and 443model ILS and other processes such as gene flow, will be necessary to better understand the 444reasons for this phylogenetic conflict. Regarding the species-delimitation, BPP gave 445consistent results even with an alternative guide tree. Thus, even in the case of phylogenetic

446uncertainty, this method seems to be a valid alternative to BFD* for species-delimitation with 447RADseq data, even if the amount of data is still an important limitation.

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4494.2 Phylogeny and systematics of Neurergus newts

450Given their limited number of species and the rather well-supported previous mitochondrial 451phylogenies, *Neurergus* newts would at first glance appear an easy target for phylogenomics. 452However, unravelling their phylogenetic history turned out to be less simple than initially 453thought. While the position of *N. strauchii* as sister-species to all other *Neurergus* is 454unambiguous, and congruent with previous studies, the relationships within the *N. crocatus* + 455*N. kaiseri* + *N. derjugini* clade are more difficult to resolve. For these taxa our analyses 456 recovered two alternative scenarios: (A) *N. crocatus* as sister-species to *N. kaiseri* + *N.* 457*derjugini* or (B) *N. kaiseri* as sister-species to *N. crocatus* + *N. derjugini*. This second 458scenario is the most likely from a biogeographic point of view, as the ranges of *N. crocatus* 459and *N. derjugini* are quite close, while *N. kaiseri* occurs further south. Thus, after the initial 460split between *N. strauchii* and the *N. crocatus* + *N. kaiseri* + *N. derjugini* clade, divergence 461 within this group could have occurred along the Zagros mountains, with N. kaiseri first 462becoming isolated in the southern part, and *N. crocatus* and *N. derjugini* being separated 463subsequently. Although it is uncertain to what extent the ecology of these newts reflects 464phylogenetic relationships, it is worth noting that *N. crocatus* and *N. derjuqni* live in similar 465habitats (mountain streams), while *N*. *kaiseri* has adapted to more pond-dwelling habits, 466 facing the more arid conditions at the southern distribution range of the whole genus 467(Steinfartz et al., 2002).

468 The RADseq data also revealed a high degree of intraspecific diversity in 469*Neurergus*. Except for *N. kaiseri*, our geographical sampling per species is limited, yet we

470uncovered a pattern of strong molecular divergence among populations, not only for *N*. 471*kaiseri* but also for *N. crocatus*. In *N. crocatus*, the individual from the north (Semdinli) 472turned out to be clearly separated from the southern ones (Shiwalok). This structure in *N*. 473*strauchii* corresponds to the two described subspecies, *N. s. strauchi* and *N. s. barani*. Also 474for *N. derjugini*, two genetically divergent subspecies (*N. d. derjugini* and *N. d.* 475*microspilotus*) are known, and previous analysis of two nuclear gene fragments failed to 476place these two lineages into a monophyletic group (Hendrix et al., 2014). The more 477informative RADseq data now strongly support the clade containing these morphologically 478very similar and geographically close lineages and indicate only shallow genetic divergences 479between them. Moreover, in *N. kaiseri*, a substantial divergence was detected between 480populations from the northern and southern parts of its range, but this divergence is not 481currently reflected in the species' taxonomy.

Using coalescent-based species-delimitation models on the RADseq data, a 483classification with at least seven *Neurergus* species was strongly supported. This would 484suggest that the species richness of the genus has so far been severely underestimated based 485on morphology, mitochondrial markers and a restricted number of nuclear genes (Hendrix et 486al., 2014; Steinfartz et al., 2002). Taxonomically, this would imply that the sampled 487subspecies of both *N. strauchii* and *N. derjugini* should all be considered distinct species, as 488well as the two main clades of *N. kaiseri*. However, simulation studies have shown that the 489MSC model does not necessarily infer true species boundaries but might rather reflect 490population structure (Sukumaran and Knowles, 2017). Empirical studies using several 491species-delimitation approaches revealed that the number of species can be overestimated in 492the presence of gene flow and/or intra-specific structure (Chan et al., 2017; Miralles and 493Vences, 2013). This is particularly relevant for the split of *N. d. derjugini* and *N. d.* 494*microspilotus*. Indeed, even though they are separated as distinct species using MSC species-495delimitation, these two subspecies only show a very shallow divergence on the concatenation 496tree. We hypothesize that this split might reflect intraspecific structure, resulting from a 497decrease in gene flow between the populations, rather than completed speciation. Recent 498range reduction, leading to strict allopatry, could be the process underlying the observed 499weak differentiation. This hypothesis agrees with the lack of support that the split of *N*. 500*derjugini* received in some of the BPP analyses, and with lack of substantial increase in the 501marginal likelihood of the models in the BFD* analysis.

502 On the contrary, the split between *N*. *s*. *strauchii* and *N*. *s*. *barani* is always 503strongly supported. These two taxa show a deep divergence estimated at 3 million years ago 504based on allozyme data (Steinfartz et al., 2002). A large-scale genetic and morphological 505study revealed that the two subspecies are separated by the Euphrate River (Pasmans et al., 5062006), which could act as a long-term barrier to gene flow and therefore promote speciation. 507Major water courses as primary reason for isolation have been identified also for other 508amphibians (Zeisset and Beebee, 2008). The analysis of morphological data from the whole 509range of *N. strauchii* has also revealed morphological differences between the two 510subspecies, particularly in a significantly higher number of yellow spots in *N*. *s*. *strauchii* 511(Pasmans et al., 2006). Thus, considering the concordant genomic and morphological 512divergence, we suggest treating *N*. *s*. *strauchii* and *N*. *s*. *barani* as distinct species, elevating 513the latter to species status: *N. strauchii* (Steindachner 1887), and *N. barani* (Öz 1994). A full 514 revision of the species complex, however, will necessitate inclusion of *N. s. munzurensis*, a 515recently described subspecies (Olgun et al., 2016) not present in our data set. As it is 516morphologically distinct from both *strauchii* and *barani*, and separated from them by large 517water courses, it might also be characterized by a relevant genetic divergence.

518 Finally, the status of *N*. *kaiseri* is more difficult to assess without additional 519analyses. MSC species-delimitation analyses strongly support that individuals from the North 520and the South of the distribution area are independent lineages. A recent study of the 521mitochondrial D-loop supports this result, as these two groups do not share any haplotype 522(Farasat et al., 2016). These results are in concordance with a recent morphological study that 523highlighted significant morphometric differentiation between the groups. However, even 524though the phylogenetic divergence between lineages seems to be high, their pairwise 525nucleotide diversity is low in comparison to that of *barani* and *strauchii*, for example. Thus, 526this seemingly high diversity in comparison to other *Neurergus* species could be an artefact 527due to the comparable larger sample size (7 individuals versus an average of 3.3 individuals 528included for the other species). However, in the BFD* analyses, the model splitting *N. kaiseri* 529alone had a higher marginal likelihood than the one splitting *N. strauchii*. Given our sampling 530of the former species, biased toward the northern clade (5 samples vs. 2 in the Southern 531clade), it is possible that we are underestimating their pairwise nucleotide diversity. 532Population-level analyses of a more extensive molecular dataset are needed to reveal whether 533there is still gene flow going on between the two lineages.

Overall, speciation in *Neurergus* appears to be driven mostly or exclusively by 535allopatric processes, with no sympatry known between species. In this respect, field 536exploration would be particularly important to detect possible geographically intermediate 537populations or even contact zones, e.g. of the two subspecies of *N. derjugini*. In the case of 538*N. kaiseri*, an in-depth look at the contact zone between the two lineages will permit 539identifying possible gene flow, or lack thereof, as well as ecological factors underlying their 540geographical separation. Finally, an extensive genomic sampling of all the *Neurergus* species 541would also be very interesting to know if they consistently display a genetic diversity

542comparable to that of *N. kaiseri*, which could then be attributed to specific features of these 543newts such as possible low dispersal abilities. Otherwise, understanding the processes 544underlying the highest genetic richness of *N. kaiseri* could reveal very interesting elements of 545their evolutionary history, such as ecological shifts, or a particular biogeographic history.

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557**Competing interest statement**

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559The authors declare that there is no competing interest.

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563**References**

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795Figure 1. Map of the sampling localities of *Neurergus* newts. Dashed lines show the whole

796range of the different taxa (distribution data from Steinfartz et al. 2002). The number of

797samples included in this study from each site is indicated in parentheses.

798

799Figure 2. Maximum Likelihood tree of Neurergus newts inferred from the concatenation of

8008,314 RAD loci (long branches have been artificially shortened). Stars at nodes indicate

801100% bootstrap support and 100% jackknife support from ca. 100,000 positions sampled.

802

803**Figure 3.** Species-tree of *Neurergus* newts inferred (a) using tetrad, with 8,314 SNPs, with 804bootstrap support from 500 replicates. When bootstrap supports differs on the tree without 805outgroup (14,765 SNPs), both values are indicated; (b) using SNAPP on the dataset without 806outgroup, with posterior probabilities of the nodes indicated.

807

808**Figure 4.** Kernel density of the proportion of gene trees supporting each topology for 100 809random draws of 1 individual per species, with dashed lines showing the average support.

810The blue curve represent the Topology B, the green one the Toplogy A and the red one an 811alternative Topology with *N. kaiseri* and *N. crocatus* as sister species.

Figure 1



Figure 2



Figure 3

0 -

0.100

0.125



0.150

proportion of gene trees

1

0.200

0.175

Table S1. Summary of the 66 dataset used to assess the effect of clustering threshold and minimum individual coverage on phylogenetic reconstruction. Concat. = concatenation of the loci; USNPs = concatenation of the unlinked SNPs (one per locus). The total length of the concatenation is given with the percentage of parsimony informative sites between parentheses, as well as the total number of unlinked SNPs. The datasets that yielded a topology that differs from the majority one (given in Figure 2) are in bold if the difference is in the deep nodes (higher than specific level) or underlined if the difference is in intra-specific nodes.

	Min. individual	Clustering threshold = 85%		Clustering threshold = 90%		Clustering threshold = 95%	
	coverage	Concat.	USNPs	Concat.	USNPs	Concat.	USNPs
	4	9,300,13 5 bp (1.51%)	66,106 SNPs	9,820,74 8 bp (1.39%)	68,962 SNPs	10,947,7 32 bp (1.11%)	74,132 SNPs
	8	5,343,79 9 bp (2.10%)	43,072 SNPs	5,540,92 4 bp (1.97%)	44,421 SNPs	5,718,39 7 bp (1.60%)	44,844 SNPs
With outgroup	10	4,214,57 8 bp (2.64%)	34,390 SNPs	4,335,43 3 bp (2.13%)	35,198 SNPs	4,271,83 8 bp (1.72%)	33,873 SNPs
(19 individuals)	14	1,892,18 5 bp (2.61%)	15,830 SNPs	1,882,12 6 bp (2.46%)	15,682 SNPs	1,583,47 2 bp (1.96%)	12,909 SNPs
	16	991,030 bp (2.81%)	8,379 SNPs	955,241 bp (2.65%)	8,052 SNPs	702,497 bp (2.09%)	5,802 SNPs
	18	157,833 bp (2.91%)	1,366 SNPs	104,396 bp (2.52%)	904 SNPs	20,951 bp (1.79%)	<u>181</u> <u>SNPs</u>
Without outgroup (17 individuals)	4	9,269,61 9 bp (1.49%)	64,449 SNPs	9,841,31 9 bp (1.39%)	67,912 SNPs	11,025,2 40 bp (1.10%)	74,201 SNPs
	8	5,337,23 0 bp (2.07%)	42,377 SNPs	5,564,65 3 bp (1.96%)	44,071 SNPs	5,759,11 3 bp (1.60%)	44,948 SNPs
	10	4,189,66 1 bp (2.83%)	33,703 SNPs	4,343,77 6 bp (2.12%)	34,863 SNPs	4,304,92 2 bp (1.72%)	33,979 SNPs
	14	1,812,60 5 bp (2.57%)	14,941 SNPs	1,835,85 4 bp (2.45%)	15,114 SNPs	1,582,49 1 bp (1.96%)	12,841 SNPs
	16	918,189 bp (2.75%)	7,651 SNPs	912,210 bp (2.65%)	7,603 SNPs	697,197 bp (2.09%)	5,729 SNPs

Figure S1. Bootstrap support for the trees inferred from the 66 alternative datasets. The average support is given for the deep nodes (relationships between species, triangles), and for the intraspecific nodes (squares) inferred from the concatenation of all the loci (blue) and only unlinked SNPs (red). The standard deviations are represented as error bars around the average values.

