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Comparison of four species-delimitation methods applied to a DNA barcode data set of insect larvae for use in routine bioassessment

Bryan P. White 1,4, Erik M. Pilgrim 2,5, Laura M. Boykin 3,6, Eric D. Stein 1,7, and Raphael D. Mazor 1,8

Abstract: Species delimitation (grouping individuals into distinct taxonomic groups) is an essential part of evolutionary, conservation, and molecular ecology. Deoxyribonucleic acid (DNA) barcodes, short fragments of the cytochrome c oxidase subunit I (COI) gene, are being used in environmental bioassessments to assign specimens to putative species, but no method for delimiting DNA barcodes into species-level entities is universally accepted. We investigated the effect of delimitation methods on outcomes of bioassessments based on DNA barcodes. We used 2 tree-construction methods (neighbor joining [NJ], maximum likelihood [ML]) and 4 classes of species-delimitation criteria (distance-based, bootstrap support, reciprocal monophyly, and coalescentbased) with a DNA barcode data set consisting of 3 genera and 2202 COI sequences. We compared species delimitations for Baetis (Ephemeroptera:Baetidae), Eukiefferiella (Diptera:Chironomidae), and Simulium (Diptera: Simuliidae) from different streams. We assessed congruence among trees and compared species abundances and estimated species richness among methods. NJ followed by use of a standard barcoding distance cutoff (2%) yielded the greatest number of putative species. All other delimitation methods yielded similar, but lower, richness. Differences in species delimitations produced by various methods might have been caused by confounding factors, such as possible parthenogenesis in Baetis and rare haplotypes in abundant species of Baetis and Simulium. Eukiefferiella presented the fewest discrepancies among delimitations. Each method can be regarded as producing a separate line of evidence contributing to the delimitation of separately evolving lineages. The increased resolution offered by DNA barcoding can yield important insights into the natural history of organisms, but the power of these observations is limited without the use of multigene and multilocus data sets. Key words: DNA barcoding, bioassessment, biomonitoring, species delimitation, coalescent theory, cytochrome oxidase I, COI, benthic macroinvertebrates, GMYC

Deoxyribonucleic acid (DNA) barcoding is a molecular taxonomic method wherein an \sim 650-base pair (bp) region of the cytochrome c oxidase subunit I (COI) mitochondrial gene is used as a species-level molecular identification marker in animals (Hebert et al. 2003). DNA barcoding has the potential to affect the field of bioassessment (also called biomonitoring) where biological measures, such as species richness and taxonomic composition, are used to draw conclusions about the health of an ecological system. Incorporation of DNA barcode methods into bioassessment programs has been evaluated theoretically (Jones 2008) and empirically (Pilgrim et al. 2011, Sweeney et al.

2011, Stein et al. 2013, 2014). The conclusion is that molecular methods might improve our view of stream diversity because they provide increased taxonomic resolution.

In most bioassessment programs, estimates of species richness are obtained via morphological identification. However, this method can be difficult and time consuming, and morphology cannot be used to differentiate cryptic species (Bickford et al. 2007) or species complexes (Hajibabaei et al. 2006). The ability to differentiate such species is important because presently unknown differences in their tolerance to pollution, reproductive timing, feeding mechanisms, or other ecological traits may provide

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clues regarding the health of a stream system (Verberk et al. 2013). Moreover, species-delimitation methods are relevant to more than the outcomes of routine bioassessment programs. They may have utility in the fields of criminal wildlife forensics (Dawnay et al. 2007), biodiversity indexing (Janzen et al. 2009), detection of fish-market replacements (Maralit et al. 2013), ecology (Valentini et al. 2009), and biosecurity (Boykin et al. 2012). In each application, routine, repeated sampling of organisms should yield consistent sets of species designations across different lab-

Use of molecular taxonomic methods, such as DNA barcoding, to identify unknown organisms or organisms impractical to identify to species level on a routine basis has led to questions related to how to delimit species based only on a DNA barcode. Rapid decreases in the cost of high-throughput Sanger sequencing and the advent of Next Generation Sequencing (NGS; Shendure and Ji 2008) have led to an exponential increase in the rate at which DNA barcodes are being uploaded to public databases. Thus far, 1.2 million COI sequences have been released on the Barcode of Life Data System (BOLD; Ratnasingham and Hebert 2007; www.barcodinglife.org), as part of the International Barcode of Life Project (iBOL, http://ibol.org/), and another 1 million remain uploaded but unreleased. This vast amount of data has created a pressing need for efficient, objective, and readily reproducible algorithms for species delimitation. Many proponents of DNA barcoding have suggested the use of measures of genetic distance to designate species. Investigators have recommended species limits based on an average genetic distance of ≥2% (see below) among individuals in different putative species (Ball et al. 2005, Zhou et al. 2009) or on a level of interspecific variation that is 10× the intraspecific variation (Hebert et al. 2004). Application of coalescent-theoretic methods (Knowles and Carstens 2007, Rodrigo et al. 2008, Zaldívar-Riverón et al. 2010, Zhang et al. 2011, Nuñez et al. 2012, Vuataz et al. 2012), the principal of genealogical sorting (Cummings et al. 2008), machine learning methods (Bertolazzi et al. 2009, Weitschek et al. 2013), a heuristic-search-strategy method (O'Meara 2010), Bayesian statistical methods (Yang and Rannala 2010, Zhang et al. 2011), and a multimethod 'tip to root' approach (Boykin et al. 2012) have been proposed as ways to increase objectivity and biological relevance of species delimitation. We evaluated 4 major classes of species-delimitation criteria (the criteria by which a 'haplotype cluster' is granted a specieslevel status): genetic distance-based (DB), bootstrap support (BSS), reciprocal monophyly (RM), and coalescentbased (CB). We applied these criteria by constructing 2 types of phylogenetic trees. First, we constructed neighbor joining trees (NJ), which implement a clustering algorithm that always finds the 'first best' (balanced minimum evolution) tree given the data set. Second, we constructed maximum likelihood trees (ML), which implement an algorithm that heuristically searches a subset of all possible

trees to find the highest log-likelihood (lnL) tree. We focused on 3 insect genera that are widely encountered in freshwater bioassessment and whose species are difficult to identify morphologically: Baetis (Ephemeroptera:Baetidae), Eukiefferiella (Diptera:Chironomidae), and Simulium (Diptera:Simuliidae). Decisions regarding species delimitation may affect bioassessment metrics. Therefore, we addressed the following questions: 1) Do different delimitation methods yield different numbers of detectable species? 2) Are differences among methods in the number of species detected associated with the number of haplotypes (i.e., are haplotype-rich species more difficult to delimit, or vice versa)? 3) Do estimates of species richness produced by different delimitation methods differ among sampling sites?

METHODS

Study site and genera

We obtained a subset (2202) of COI sequences from Baetis, Eukiefferiella, and Simulium from a bioassessment study of 5 streams in the Los Angeles, California (USA), area (Stein et al. 2014). Benthic macroinvertebrate samples were taken from 2 reaches at each stream. Expert taxonomists identified specimens morphologically using a standard level of taxonomic effort. These experts identified 3 distinct species of Baetis (Baetis tricaudatus, Baetis adonis, and a 3rd unknown, but recognizably distinct, species Baetis sp. CA), whereas they identified Simulium and Eukiefferiella species only to genus. We treated data for each genus separately. The sequences used in our study are publicly available under the BOLD projects CFWIA through CFWIJ (see Table S1 for a complete list of BOLD sample identification codes and GenBank accessions).

Sequence data and haplotype collapsing

We selected closely related genera as outgroups for each data set (Table 1). We used a minimum sequence length requirement of 500 base pairs (bp) to reduce uncertainty during NJ and phylogenetic analyses. We translated sequences to amino acids in MEGA (version 5.1; Tamura et al. 2011) and aligned them in MUSCLE (version 3.8.31; Edgar 2004). The final alignment was corrected manually so that sequences lacked gaps and consisted of an uninterrupted open-reading-frame, which led us to conclude that no insertions, deletions, or pseudogenes were present in the data sets. Following alignment, we used an open-source, custom Perl script developed for this analysis (dnab_collapse.pl, https://github.com/bpwhite /bioinformatics-toolbox) to reduce the number of individual sequences and, thus, the computational requirements for each analysis. Important information, such as location, haplotype identification, and taxonomic identification were preserved. The end result of this process was sequences

Table 1. Number of species identified in each genus with each species-delimitation method. ML = maximum likelihood, NJ = neighbor-joining, CB = coalescent-based, RM = reciprocal monophyly, BSS = bootstrap support.

Taxon	ML+CB	ML+RM	NJ+BSS	NJ+DB
Baetis	3	4	4	5
Eukiefferiella	10	10	10	9
Simulium	8	7	7	10

that were either unique haplotypes or haplotypes that were present at ≥2 sites. Each remaining sequence was automatically annotated with the abundance of that haplotype at each location so that information about the diversity and abundance of haplotype clusters could be garnered quickly.

Intraspecific-variation method (NJ+DB)

The DB criterion is based on use of an a priori genetic distance threshold as the cutoff for deciding whether 2 individuals are members of the same species. This criterion is predicated on the idea that intraspecific genetic variation is small relative to interspecific variation. For example, if the distance cutoff is 2% (Herbert et al. 2003, Meyer and Paulay 2005, Rivera and Currie 2009, Sweeney et al. 2011), and the calculated genetic distance between individual A and B is 2.5%, then the 2 individuals are assigned to different species. Variations on this method include use of average intraspecific distances (Hebert et al. 2004, Zhou et al. 2009, 2011) or variable thresholds depending on the taxa (Sweeney et al. 2011). The DB method typically has been applied to NJ trees computed with the algorithm of Saitou and Nei (1987) and the Kimura-2-parameter (K2P; Kimura 1980) measure of genetic distance.

We applied a DB criterion by calculating the nearestneighbor distance (smallest interspecific distance) between haplotype clusters using the Species Delimitation (version 1.04; Masters et al. 2011) plugin for Geneious (version 5.6.5; Biomatters, http://www.geneious.com/). Two haplotype clusters that contained a pair of nearest neighbors with >2% K2P distance from each other were considered different putative species.

Statistical methods (NJ+BSS and ML+RM)

BSS is the proportion of bootstrap replicates in which particular sequences clustered together when the NJ algorithm is applied (Felsenstein 1985). For example, if a node achieves 95% bootstrap support, then that node and all of its children were grouped together in 95% of the bootstrap replicates. This method has been used in large-scale DNA barcoding studies by Zhou et al. (2009, 2011), Lakra et al. (2011), and Mecklenburg et al. (2011).

We implemented the NJ+BSS method using 1000 bootstrap replicates in MEGA. We used K2P distance because it is considered 'standard' in DNA barcode studies (Hebert

et al. 2003, 2004, Zhou et al. 2009, 2011, Oceguera-Figueroa et al. 2010, Sweeney et al. 2011). However, Srivathsan and Meier (2011) and Collins et al. (2012) recently suggested that K2P is rarely the best nucleotide model for COI-only data sets. We applied the BSS criterion to the bootstrapped NJ tree to define putative species based on a bootstrap support cutoff of 95%.

RM is a statistical approach based on the principal that individuals from different species will separate consistently into distinct monophyletic clades with >95% statistical support. RM can be applied to either maximum parsimony (MP) or ML trees, but a more rigorous test of monophyly (not done here) requires that the observed branching pattern be tested against a random branching pattern (Rosenberg 2007).

We implemented the ML+RM method by first identifying the optimal nucleotide model for each data set with jModelTest (version 0.1.1; Posada 2008). Following nucleotide model selection, we constructed ML phylogenetic trees using a Bayesian phylogenetic program, BEAST (version 1.7.4; Drummond et al. 2012) with a coalescent-tree prior and 3 different molecular clock models: strict, relaxed lognormal, and relaxed exponential. Each clock model began with a normally distributed clock rate with a mean of 0.02 substitutions/million y (s/my) (Brown et al. 1979) and a standard deviation of 0.005 s/my. We ran a Monte Carlo Markov Chain (MCMC) simulation for 10 million steps and sampled trees from the MCMC at 1000-step intervals. We checked parameter values for effective sample sizes (ESS) >200 and convergence by plotting marginal probabilities in Tracer (version 1.5; Drummond et al. 2012). We discarded the first 20% of trees sampled as burn-ins. We loaded the remaining 8001 trees into TreeAnnotator (version 1.7.4; Drummond et al. 2012) for construction of the maximum clade credibility (MCC) tree and calculation of posterior probabilities and node ages. After trees were annotated, we used a Bayes factor (BF) analysis to test whether the data were clock-like (Drummond and Rambaut 2007). In this analysis, the marginal likelihoods of each tree are estimated using the harmonic mean, and the possible improvement of one model over another is assessed by dividing their marginal likelihoods when those models differ by only 1 parameter (in this case, the clock model). The ratio of this division is the BF. An improvement of one model over another is considered significant if BF > 2. Following selection of the best tree clock model, the MCC was annotated in FigTree (version 1.4; http://tree.bio.ed.ac.uk/soft ware/figtree/).

Modeling method (ML+CB)

CB is a modeling approach derived from population genetics and is based on the principal that individuals that possess different species-level coalescent points (hypothetical ancestors of haplotypes, alleles, or species from which point all current members of a population were descended) are members of different species. Here we consider the COI gene tree to be analogous to the species tree, but in many cases, gene trees do not match species trees (Liu and Pearl 2007). Analysis of multiple genes typically is required to obtain an accurate species coalescent point.

A custom R script was created (dnab_coalesce.r, also available from: https://github.com/bpwhite/bioinformatics -toolbox) to run the CB species-delimitation analysis. This script makes use of the splits package (http://r-forge.r -project.org/projects/splits/), and imports the resultant MCC trees for each data set into the General Mixed Yule Coalescent (GMYC) function (gmyc). The gmyc function finds the ML threshold for the transition threshold from a Yule process (interspecific branching rates) to a coalescent process (intraspecific branching rates) (Pons et al. 2006, Fontaneto et al. 2007, Knowles and Carstens 2007, Monaghan et al. 2009, Nuñez et al. 2012, Vuataz et al. 2012). A likelihood ratio test is automatically performed to compare the coalescent model to a Yule model of evolution, and if the ratio results in a p-value <0.05, the coalescent model is accepted over the Yule model and the putative species entities can be considered statistically significant. We ran the GMYC model test for a single ML threshold and multiple ML thresholds, wherein the threshold was allowed to vary across lineages. We compared the results of the 2 models with a χ^2 goodness-of-fit test (also available in the splits package under the function compare). The multiple threshold test was considered an improvement over the single threshold test if the data fit that model significantly better (p < 0.05). After model selection, the dnab_coalesce.r script outputs the resultant species delimitation using the spec.list function into a commaseparated-value (CSV) format for import into other programs.

Species richness and abundance calculations

We assigned 4 putative species identifications (one for each method) to each individual sequence and used a χ^2 goodness-of-fit test to assess whether species richness was affected by the method used. We assessed the effects of species-delimitation methods on stream species richness by summing the number of putative species encountered in each stream for each method. We assessed where shifts in the abundance of species might occur by summing the number of individuals given a particular species identification for each method.

RESULTS

Data reduction of the 3 data sets decreased the numbers of sequences from 951 to 201 for Baetis, 906 to 389 for Simulium, and 345 to 32 for Eukiefferiella. The number of putative species did not differ among delimitation methods for any genus ($\chi^2_{\text{Table }2} = 1.327$, p > 0.05).

Intraspecific variation method (NJ+DB)

The NJ+DB method delimited more putative Baetis species than all other methods (Table 1) because it split Baetis 1 into 2 species (1 and 2) (Fig. 1A, B). The NJ+DB method delimited fewer putative Eukiefferiella species than all other methods (Table 1) because it lumped Eukiefferiella 5 and 6 into 1 species. The genetic distance between Eukiefferiella 5 and 6 was 1.9%, thus species 6 missed the cutoff by 0.1% (Fig. 2A, B). The NJ+DB method delimited more putative Simulium species than all other methods (Table 1) because it split Simulium 1 into 2 species (1 and 2) and Simulium 9 into 2 species (9 and 10) (Fig. 3A, B).

Statistical methods (NJ+BSS and ML+RM)

Both statistics-based methods resulted in identical species delimitations in all 3 genera, but the use of ML tree construction methods increased support values for many nodes over the bootstrap support values (Figs 1-3).

Modeling method (ML+CB)

Each data set had a different nucleotide model. Hasegawa-Kishino-Yano + gamma [HKY+G] was selected for Baetis, general time reversible + gamma (GTR+G) was selected for Simulium, and general time reversible + invariant + gamma (GTR+I+G) was selected for Eukiefferiella (Table 2). The lognormal and exponential relaxed clock models were not significant improvements over the strict clock model for any genus (BF < 2 in all cases), so the strict clock model was used for both Baetis and Eukiefferiella. In the case of Simulium, negative branch lengths in the strict clock tree made the application of the GMYC model impossible. We used the lognormal clock tree instead because it had only a slightly faster mean rate than the strict clock tree (strict: 0.199 vs lognormal: 0.244). The Simulium MCMC may have been undersampled because the ESS for the likelihood parameter was <200, whereas the coalescent parameter was >200 (Table 2). For all 3 genera, the GMYC model was selected over the null model of evolution (Table 2), and the multiple ML threshold model was not a significant improvement over the single threshold model (Table 2). The ML+CB method produced the same number of species as the ML+RM and NJ+BSS methods, but the designations of those species were different, for example, Baetis 3 was not split into Baetis 3 and 4 (Fig. 1A, B), whereas it was in the other 3 methods. Moreover, Simulium 8 and 9 were identified by the ML+CB method, but Simulium 10 was not (Fig. 3A, B).

Shifts in species abundance and richness

The only difference in species abundances among delimitation methods was for Baetis 1 and 2. The NJ+DB method yielded 2 species consisting of 364 and 429 indi-

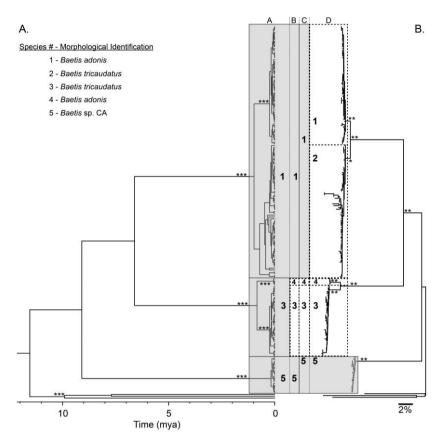


Figure 1. A.—Maximum likelihood, strict clock tree of 201 Baetis cytochrome c oxidase subunit I (COI) sequences computed using BEAST with a coalescent tree prior and the Hasegawa-Kishino-Yano + gamma (HKY+G) nucleotide model. Species entities were delimited using the gmyc function of the splits package 1.0-14. B.—Neighbor-joining tree computed using MEGA 5.05 and the Kimura 2-parameter distance nucleotide model. Putative species are numbered in order of appearance on the tree and highlighted in shaded grey boxes. Putative species that undergo splitting are indicated by dashed boxes. A = ML+CB, B = ML+RM, C = NJ+BSS, D = NJ+DB, = 0.80–0.94 node support, ** = 0.95–0.99 node support, *** = 1.00 node support.

viduals, whereas the other methods yielded 1 species with 793 individuals. Minor differences in species abundances among methods were present in all 3 genera, but these differences were limited mostly to the presence or absence of a few rare haplotypes. Species richness differed consistently among methods and sites (Fig. 4). When estimates differed among methods, the NJ+DB method typically produced higher species richness than the other methods. In all cases the RM and BSS method produced identical abundance and richness estimates.

DISCUSSION

Four species-delimitation methods applied to a data set of 2202 COI sequences from 3 genera of insect larvae from southern California yielded similar estimates of species richness. Where standard morphological identification effort yielded 5 distinct taxa (Baetis adonis, Baetis tricaudatus, Baetis sp. CA, Simulium, and Eukiefferiella), DNA barcodes yielded 19 to 25 putative species, a 4× increase in resolution over the standard level of identification. The differences among species delimitations, although not statistically significant, tended to be associated with abundant and diverse taxa. This result suggests that the uncertainty associated with species delimitations derived from DNA barcoding does not arise from the algorithm used, but is a byproduct of the inherent limitations of COI as a species-level phylogenetic marker. Differences among delimitation methods are not likely to result in large changes in bioassessment metric scores based on taxon richness.

The absence of noticeable differences in species abundances under different delimitation methods (except Baetis 1 and 2) suggests that richness metrics that take abundances into account might be unaffected by different delimitation methods, whereas presence-absence-type richness metrics may be more directly influenced by even subtle shifts in species designations (e.g., Simulium 2 and 10 exist only under the NJ+DB method and have extremely low abundances [1 and 2 individuals, respectively]). How-

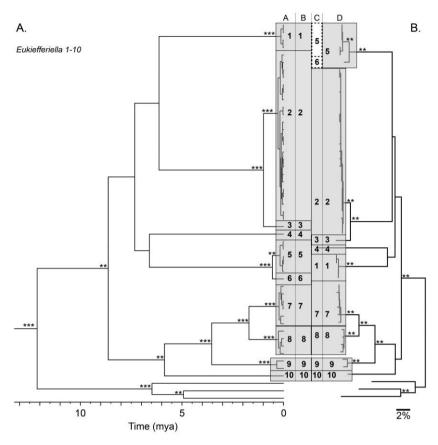


Figure 2. A.—Maximum likelihood, strict clock tree of 32 *Eukiefferiella* cytochrome *c* oxidase subunit I (COI) sequences computed with the GTR+I+G nucleotide model. B.—Neighbor-joining tree. See Fig. 1A, B for details of species delimitation, tree construction, and figure conventions.

ever, even those small differences may not be large enough to significantly affect bioassessment metrics beyond the greater taxonomic resolution already provided by DNA barcoding.

Species delimitations that differed among methods tended to be associated with high-abundance species with many different COI haplotypes, a potentially important trend warranting further examination. For example, putative species that experienced splits under the NJ+DB method (Baetis 1-2, 3-4, and Simulium 1-2, 9-10) were usually very abundant (>30 individuals encountered). Large populations tend to be less prone to chance events like genetic drift that eliminate rare haplotypes (given enough time), and it seems plausible that very abundant species will maintain more rare haplotypes in their gene pools than will less abundant species. This idea received support from Bergsten et al. (2012), who concluded that the uncertainty of species identifications for Agabini diving beetles increased significantly when sampled over increasing geographic distances. They found that a sample size of 70 individuals was necessary to capture 95% of intraspecific diversity. We observed a pattern in Baetis sp. 1 and 2, which made up 83% of Baetis encountered (793 individ-

uals), of increased diversity with increased intraspecific sampling effort, a result that further supports the hypothesis of Bergsten et al. (2012). In our data set, Baetis 1 and 2 had an average intraspecific K2P distance of 1.6% and a maximum pairwise K2P distance of 5.1% when lumped together according to the NJ+BS, ML+RM, and ML+CB criteria. The less frequently encountered Baetis species (3, 4, and 5), which made up only 17% of Baetis encountered (158 individuals), did not exhibit intraspecific pairwise distances >1%. These differences in sampling effort reflect natural abundances of these species and not targeted sampling effort toward one species or another. In contrast, we also found many individuals in Baetis 1 and 2 that were separated by great geographic distances but shared identical haplotypes. For example, 8 individuals found in Conejo Creek had haplotypes identical to 50 individuals in Big Tujunga Wash. The 2 streams are separated by a geographic distance of 64 km and an elevation difference of 335 m (see Stein et al. 2014 for stream descriptions). We did not see more than 1 or 2 instances of shared haplotypes over great distances in Simulium or Eukiefferiella.

The multimodal pattern of genetic variation within the *Baetis* 1 and 2 complex consists of a mixture of low

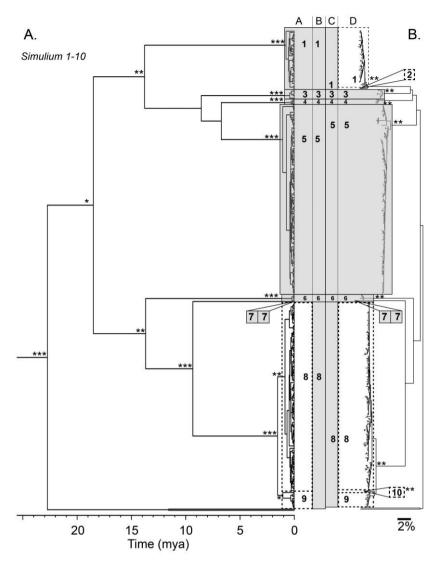


Figure 3. A.—Maximum likelihood, lognormal clock tree of 389 Simulium cytochrome c oxidase subunit I (COI) sequences computed with the GTR+G nucleotide model. B.—Neighbor-joining tree. See Fig. 1A, B for details of species delimitation, tree construction, and figure conventions.

and high diversity over broad distances. Such a pattern might be explained if these species were parthenogenetic. Many mayfly species exhibit parthenogenesis (Bergman and Hilsenhoff 1978, McCafferty and Morihara 1979, Funk et al. 2006, 2008), and Funk et al. (2010) suggested that most, if not all, mayflies may be facultatively parthenogenetic. Females may reproduce parthenogenetically, but offspring can be either male or female and can readily revert to sexual reproduction. Vanoverbeke and De Meester (1997) found no relationship between geographic distance and genetic distance in parthenogenetically reproducing populations of cladoceran branchiopods (Daphnia magna), a result similar to our finding of shared haplotypes over large distances (64 km). However, we cannot rule out the possibility that this pattern is a by-product of normal metazoan mitochondrial inheritance and that shared haplotypes are a consequence of evolutionarily recent dispersal events.

The increased taxonomic resolution offered by DNA barcoding could be used to understand better the life histories of the seemingly cryptic Baetis observed in our study, which might in turn help researchers use traits-based approaches to ecology and improve future bioassessment tools (Verberk et al. 2013). In the case of Baetis spp. in southern California, the addition of nuclear loci to DNA barcode data sets might help researchers distinguish between normal and parthenogenetic modes of inheritance (Buckley et al. 2008). When systematists undertake taxonomic revisions of morphologically cryptic species in light of molecular data (e.g., in the Atyaephyra genus of freshwater shrimp; Christodoulou et al. 2012), specific life-

(BOLD) sample identification numbers (IDs), GenBank accession numbers for each outgroup, nucleotide model used, tree log likelihood (lnL), effective sample size (ESS) of tree lnL and coalescent BEAST parameters, null (Yule) model lnL from gnyc function, General Mixed Yule Coalescent (GMYC) model lnL, p-value of the χ^2 goodness-of-fit test between the null and GMYC model, p-value of the comparison between the single and multiple (multi) GMYC threshold, and number of putative species detected under the GMYC model (ML entities). Mya = million years ago. Taxa used, number of cytochrome c oxidase subunit I (COI) sequences, number of haplotypes for each taxa, outgroups used for each taxa, Barcode of Life Database

Baetis	906	sequences mapiotypes	Outgroups	BOLD sample ID	accession number	Nucleotide model	Tree lnL	ESS of tree lnL	ESS of coalescent	model	model InL	$\operatorname{GMYC} p$	Multi P	Threshold (mya)	ML entities
		104	Fallceon sp. Centroptilum triangulifer Callibaetis	10-SCCWRP-6041 09LJ17 09NBMAY-0053	JN297857 HM423544 JQ663249	HKY+G	-3987.374	203.5	963.9	1195.81	1220.916 <<0.0001	<<0.0001	0.884	0.973	e
Simulium	951	304	ferrugineus Prosimulium mixtum Prosimulium	08-SWRC-1082 FJ524559	JF287830 FJ524559	GTR+G	-6018.239	22.5	368.4	2258.584	2312.053	<<0.0001	0.917	1.063	
Eukiefferiella	345	32	travisi Chironomus	ChiBlank_F05	HM137935	GTR+I+G	-3826.534	857.1	3744.5	128.73	145.469	<<0.0001	0.999	0.333	10
			грания Сhironomus Fiionsis	JQ350720	JQ350720										
			Chironomus jacksoni	CAUS016-09	1										

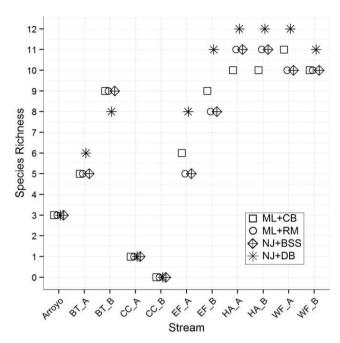


Figure 4. Differences in species richness observed at different southern California stream reaches.

history traits, such as parthenogenesis, could be included in descriptions, transferred to traits databases, and associated with DNA barcodes.

Coalescent models take into account the natural birth and death processes of populations, avoid the use of a priori distance cutoffs (e.g., the 2% cutoff), and provide a statistical framework for testing species delimitations. Furthermore, the GMYC model provides estimates of the times of transitions from inter- to intraspecies branching patterns and effectively links the fields of population genetics and phylogenetics. However, multiple genes from both nuclear and mitochondrial DNA are required to obtain a robust estimate of the species coalescence (Heled and Drummond 2010, Fujita et al. 2012). If we regard a species as any separately evolving metapopulation of lineages (de Queiroz 2007), the ML+CB method stands beside distance and monophyly-based methods as a separate and unique line of evidence in the diagnosis of distinct lineages. Thus, combining the results of several species-delimitation methods might allow researchers to draw confident conclusions when diagnosing a lineage (as in Boykin et al. 2012).

Differentiating between cases of rare, divergent haplotypes within a species and cryptic species is one of the greatest challenges for users of DNA barcodes as species-level markers. This challenge probably will be overcome only by using multigene and multilocus reference libraries. A combination of mitochondrial genes (e.g., COI, CYTB, and 16S) and nuclear genes (e.g., 18S, 28S, ITS1, and ITS2) would allow robust estimates of species coalescence and improved phylogenetic resolution. Use of

multigene data sets in routine bioassessments might be cost prohibitive because linking multiple genes to a single voucher specimen requires sorting and molecular tagging of each individual voucher specimen. As metabarcoding of environmental samples becomes more prevalent (Carew et al. 2013, Ji et al. 2013), sorting of individual organisms might become less problematic. One solution might be to maintain multigene reference libraries of local taxa identified with rigorous species-delimitation methods, but to use single-gene methods during routine field sampling (e.g., COI- or 16S-based DNA barcoding coupled with distancebased algorithms). Multigene reference libraries would allow researchers to know a priori when instances of mitochondrial introgression (incorrect lumping) or ancestral polymorphisms (incorrect splitting) might produce spurious single-gene delimitations so that degrees of confidence could be assigned to molecularly derived species delimi-

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