

1

1 For Consideration in Systematic Biology

2

3 **Evaluating fossil calibrations for dating phylogenies in light of rates of molecular evolution: a comparison**
4 **of three approaches**

5

6 Vimoksalehi Lukoschek^{1,2}, J. Scott Keogh³ and John C. Avise¹

7

8 ¹*Department of Ecology and Evolutionary Biology, University of California at Irvine, Irvine, CA, 92697, USA*

9 ²ARC Centre of Excellence for Coral Reef Studies, James Cook University, Townsville, QLD 4811, Australia.

10

11 ³*Division of Evolution, Ecology and Genetics, Research School of Biology, The Australian National University,*
12 *Canberra, ACT 0200, Australia*

13

14 Corresponding Author:

15 Vimoksalehi Lukoschek

16 ARC Centre of Excellence for Coral Reef Studies,

17 James Cook University,

18 Townsville, QLD, 4811, Australia

19 Phone: +61-7-47816294

20 Email: vimoksalehi.lukoschek@jcu.edu.au

21

22 **Running head:** Evaluating fossil calibrations

2

23Abstract

24Evolutionary and biogeographic studies increasingly rely on calibrated molecular clocks to date key events.
25While there has been significant recent progress in development of the techniques used for molecular dating,
26many issues remain. In particular, controversies abound over the appropriate use and placement of fossils
27for calibrating molecular clocks. Several methods have been proposed for evaluating candidate fossils,
28however, few studies have compared the results obtained by different approaches. Moreover, no previous
29study has incorporated the effects of nucleotide saturation from different data types in the evaluation of
30candidate fossils. In order to address these issues, we compared three approaches for evaluating fossil
31calibrations: the single-fossil cross-validation method of Near et al. (2005); the empirical fossil coverage
32method of Marshall (2008); and the Bayesian multi-calibration method of Sanders and Lee (2007), and
33explicitly incorporate the effects of data type (nuclear vs. mitochondrial DNA) for identifying the most
34reliable or congruent fossil calibrations. We used advanced (Caenophidian) snakes as a case study however
35our results are applicable to any taxonomic group with multiple candidate fossils, provided appropriate
36taxon sampling and sufficient molecular sequence data are available. We found that data type strongly
37influenced which fossil calibrations were identified as outliers, regardless of which method was used.
38Despite the use of complex partitioned models of sequence evolution and multiple calibrations throughout
39the tree, saturation severely compressed basal branch lengths obtained from mitochondrial DNA compared
40with nuclear DNA. The effects of mitochondrial saturation were not ameliorated by analysing a combined
41nuclear and mitochondrial dataset. While removing the third codon positions from the mitochondrial coding
42regions did not ameliorate saturation effects in the single-fossil cross-validations, it did in the Bayesian
43multi-calibration analyses. Saturation significantly influenced the fossils that were selected as most reliable
44for all three methods evaluated. Our findings highlight the need to critically evaluate the fossils selected by
45data with different rates of nucleotide substitution and how data with different evolutionary rates affect the
46results of each method for evaluating fossils. Our empirical evaluation demonstrates that the advantages of
47using multiple independent fossil calibrations significantly outweigh any disadvantages.

48

49**Keywords:** Bayesian dating, fossil calibrations, cross-validation, nucleotide saturation, molecular clock,

52 Ideally, molecular clock calibrations are obtained from accurately dated fossils that can be assigned to nodes
53 with high phylogenetic precision (Graur and Martin 2004), but reality is generally far from this ideal
54 because of a number of important problems. The incomplete and imperfect nature of the fossil record means
55 that fossils necessarily only provide evidence for the minimum age of a clade. Many clades will be
56 considerably older than the oldest known fossil, thus nodes may be constrained to erroneously young ages
57 (Benton and Ayala 2003; Donoghue and Benton 2007; Marshall 2008). Incorrect fossil dates also arise from
58 experimental errors in radiometric dating of fossil-bearing rocks or incorrectly assigning fossils to a specific
59 stratum. In addition, misinterpreted character state changes can result in the taxonomic misidentification of
60 fossils or their incorrect placement on the phylogeny (Lee 1999). Ideally, a fossil would date the divergence
61 of two descendant lineages from a common ancestor. In reality, however, fossils rarely represent specific
62 nodes, but rather points along a branch (Lee 1999; Conroy and van Tuinen 2003). Thus, while a fossil may
63 appear to be ancestral to a clade, it is impossible to determine how much earlier the fossil existed than the
64 clade's common ancestor. Fossils also may be incorrectly assigned to the crown rather than the stem of a
65 clade (Doyle and Donoghue 1993; Magallon and Sanderson 2001). The most useful fossils are, therefore,
66 geologically well-dated, preserved with sufficient morphological characters to be accurately placed on a
67 phylogenetic tree, and temporally close to an extant node rather than buried within a stem lineage (van
68 Tuinen and Dyke 2004). However, the fossil records of many, if not most, taxonomic groups fall far short
69 of these criteria. As such, several methods have been developed for evaluating candidate fossil calibrations
70 in order to: determine their internal consistency and identify outliers (Near et al. 2005); identify lineages
71 with the best fossil coverage and identify outliers (Marshall 2008); and evaluate alternative placements of
72 fossils (Rutschman et al. 2007; Sanders and Lee 2007).

74 However, fossil calibrations are not the only difficulty in molecular dating. Other factors also contribute to
75 inaccurately calibrated molecular clocks including: incorrectly specified models of evolution (Brandley et al.
76 2011); inappropriate modelling of rate heterogeneity among lineages (Sanderson 1997; Rambaut and
77 Bromham 1998; Drummond et al. 2006); and unbalanced taxon sampling potentially resulting in node

78density artefacts (Hugall and Lee 2007). In addition, choice of genetic data or gene region can strongly
79affect estimated divergences (Benton and Ayala 2003). For example, in rapidly evolving genes, such as
80mitochondrial DNA, saturation has been shown to have the effect of compressing basal branches and
81artificially pushing shallow nodes towards basal nodes, resulting in overestimated divergence dates (Hugall
82and Lee 2004; Townsend et al. 2004; Hugall et al. 2007; Phillips 2009). However, the nature of the bias is
83complicated. For example, underestimating the true rate of hidden substitution results in tree compression:
84however, if the rate of hidden substitutions were to be overestimated, the reverse would be true. These
85effects are further complicated by the calibration placement. For example, if only deep splits are calibrated,
86then recent nodes will be biased to be younger under tree extension and older under tree compression.
87Slowly evolving genes, as are typical for nuclear DNA, are less prone to such saturation effects, however
88nuclear DNA data are not completely immune to these issues; problems of saturation also can emerge for
89slowly evolving nuclear loci if deeper divergences are being investigated. More importantly, while the
90effects of saturation have been documented for estimating divergence times (Hugall and Lee 2004;
91Townsend et al. 2004; Hugall et al. 2007; Phillips 2009; Brandley et al. 2011), the effects of saturation on
92different approaches for evaluating candidate fossil calibrations have yet to be explored.

93

94Caenophidia (“advanced snakes” comprising acrochordids, elapids, viperids and colubrids) is a group with a
95controversial fossil record. Indeed, recent papers using calibrated molecular clocks to date divergences
96among advanced snake clades highlight the extent of controversy about the placements of certain fossils
97(Wuster et al. 2007; Sanders and Lee 2008; Sanders et al. 2008; Wuster et al. 2008; Kelly et al. 2009). In
98part this controversy exists because of the relatively poor nature of the snake fossil record. Well preserved
99and relatively complete caenophidian fossils date back no further than the Miocene (Rage 1984) and often
100belong to extant genera (Rage 1988; Szyndlar and Rage 1990, 1999), thus are of little value as calibration
101points for most studies. Earlier caenophidian fossils mostly comprise isolated vertebrae, the taxonomic
102affinities of which have been strongly debated (McDowell 1987; Rage 1987). Perhaps the most
103controversial calibrations concern the origin of caenophidian snakes themselves, which has been assigned
104dates of 38 (34-48) Myr (Sanders and Lee 2008; Kelly et al. 2009); 57 (47-140) Myr (Wuster et al. 2008);
105and > 65 Myr (Noonan and Chippindale 2006a,b), based on different interpretations of the fossil record

106(Table 1). As such, very different dates have been used to calibrate the caenophidian molecular clock (Nagy
107et al. 2003; Guicking et al. 2006; Burbrink and Lawson 2007; Wuster et al. 2007; Alfaro et al. 2008; Sanders
108and Lee 2008; Wuster et al. 2008; Kelly et al. 2009).

109

110In this paper we use advanced snakes as a test case to compare three previously published methods for
111evaluating fossil calibrations: the single-fossil cross-validation method of Near et al. (2005), the empirical
112fossil coverage method of Marshall (2008), and the Bayesian multi-calibration method of Sanders and Lee
113(2007), and explicitly evaluate the effects of nucleotide saturation on the results of each method. Briefly,
114the single-fossil cross-validation approach (Near et al. 2005) evaluates candidate fossils, including the
115alternative ages or placements of fossils at some calibrated nodes, with the aim of identifying a number of
116plausible reliable calibration sets. The approach of Marshall (2008) aims to identify candidate calibrations
117with the best fossil coverage and then tests whether these fossils are potential outliers. Finally, the Bayesian
118multi-calibration approach evaluates one or more alternative calibrations in a set by comparing the Bayesian
119prior and posterior probabilities at fossil-calibrated nodes (Sanders and Lee 2007). We explicitly evaluate
120the effects of using sequence data with different rates of molecular evolution on the best fossils identified by
121each method using the same mitochondrial and nuclear sequence dataset (each with identical taxon
122sampling) for each method. In addition, we evaluate whether saturation effects can be ameliorated by 1)
123removing the third codon position of the mitochondrial coding regions and 2) analysing a combined nuclear
124and mitochondrial dataset. Our study focused on testing alternative placements or ages of controversial
125fossil calibrations (as is typical for groups with poor fossil records); however, our approach is relevant for
126any situation where numerous candidate fossil calibrations exist.

127

128MATERIALS AND METHODS

129*Fossil Calibrations, Taxon sampling, Molecular data, Convergence Diagnostics and Saturation Plots*

130Colubroid classification is in flux (Vidal et al. 2007). We use the traditional colubroid classification as
131comprising viperids, elapids, and colubrids, including colubrid subfamilies recently elevated to higher
132taxonomic ranks (McDowell 1987; Rage 1987; Lawson et al. 2005). Forty eight taxa (40 caenophidian and
133eight henophidian taxa) were chosen based on the availability of nuclear and mitochondrial sequences

10

134(Appendix 1) and to appropriately span the various fossil calibrations tested. We specifically selected fossil
135calibrations that often have been used to date recent caenophidian divergences (Nagy et al. 2003; Guicking
136et al. 2006; Burbrink and Lawson 2007; Wuster et al. 2007, 2008; Alfaro et al. 2008; Sanders and Lee 2008;
137Kelly et al. 2009) and for which we could construct nuclear and mitochondrial datasets with appropriate
138taxon sampling. Details of the fossil calibrations evaluated are given in Table 1. We constructed nuclear
139and mitochondrial datasets, each with identical taxon sampling, using >100 novel sequences generated for
140this study and published sequences obtained from GenBank (Appendix 1). The mitochondrial data
141comprised 16S rRNA (454 bp), ND4 (672 bp), and cytochrome *b* (1095 bp) and the nuclear data comprised
142the oocyte maturation factor gene (*c-mos* – 864 bp), and the recombination activating gene 1 (RAG-1 – 2400
143bp). Novel cytochrome *b*, 16S rRNA and ND4 fragments were amplified and sequenced using the primers
144published in Lukoschek and Keogh (2006), Palumbi (1996) and Forstner et al. (1995) respectively and the
145protocols of Lukoschek and Keogh (2006) and Lukoschek et al. (2007). Amplifications of RAG-1 and *c-*
146*mos* used the primers and protocols of Groth and Barrowclough (1999) and Saint et al. (1998). Newly
147generated sequences were submitted to GenBank (Appendix 1). For some taxa mitochondrial fragments
148and/or nuclear genes were concatenated from two individuals or two congeneric species to minimise the
149amount of missing sequence data, in which case the highest common taxon name was assigned (Appendix
1501). Sequences were edited in SeqMan (Lasergene v.6, DNASTAR, Inc.), aligned with Clustal W2 (default
151parameters) (Labarga et al. 2007) and visually refined. Following alignment, coding region sequences were
152translated into amino acid sequences in MacClade v.4.06 (Sinauer Inc.) using the vertebrate mitochondrial
153and nuclear genetic codes as appropriate. No premature stop codons were observed, so we are confident that
154the mitochondrial sequences obtained were mitochondrial in origin and that the nuclear genes were not non-
155functional nuclear copies (pseudogenes). Saturation plots comparing uncorrected ‘p’ genetic distances with
156General Time Reversible plus invariant plus gamma (GTRig) distances were constructed for the nuclear and
157mitochondrial datasets. In order to evaluate saturation in each of the mitochondrial codon positions, we also
158constructed saturation plots for the first, second and third codon positions of the ND4 and cytochrome *b*
159genes.

160

161The best-fit models of molecular evolution for the nuclear and mitochondrial datasets were selected based

162on Akaike Information Criteria (AIC) implemented in ModelTest 3.06 (Posada and Crandall 1998) using
163model scores (-lnL) obtained from PAUP* (Swofford 2000). We evaluated alternative partitioning
164strategies using a modified version of the Akaike information criterion for small sample sizes (AIC_c) and
165Bayesian information criterion (BIC) (McGuire et al. 2007). AIC_c and BIC values incorporate a penalty for
166increasing the number of parameters in the model, thus potentially avoiding problems with model over-
167parameterisation. Three partitioning strategies were evaluated for the mitochondrial (mtCode, mtRNA;
168mtCode1+2, mtCode3, mtRNA; mtCode1, mtCode2, mtCode3, mtRNA) and nuclear data (nDNA;
169nDNA1+2, nDNA3; nDNA1, nDNA2, nDNA3). Bayesian analyses (four incrementally heated chains run
170for 2,000,000 generations sampled every 100th generation with all substitution parameters and rates allowed
171to vary across partitions) were conducted in MrBayes (Ronquist and Huelsenbeck 2003) and used to
172evaluate combinations of character partition and evolutionary model. AIC_c and BIC values were calculated
173using the equations of McGuire et al. (2007, page 841). AIC_c and BIC criteria selected the same optimal
174partitions as follows: mitochondrial - mtCode1-GTRig, mtCode2-GTRig, mtCode3-GTRig, mtRNA-GTRig;
175mtDNA excluding third codon positions (mtDNA3rdExcl) - mtCode1-GTRig, mtCode2-GTRig, mtRNA-
176GTRig; and nuclear – nDNA1-GTRig, nDNA2-GTRig, nDNA3-GTRig with model parameters allowed to
177vary independently across partitions. However, MrBayes returned unrealistic estimates of alpha for the
178nDNA1 gamma distribution of rate heterogeneity (66.74 ± 4006.05) so we used the next best nDNA model
179(nDNA1+2-GTRig, nDNA3-GTRig) and the best mtDNA model for all Bayesian analyses (BEAST and
180MrBayes). We also conducted extensive preliminary analyses of all three methods using a combined nDNA
181+ mtDNA dataset, but the results were virtually identical to those obtained for the mtDNA data alone, so we
182do not present the results of the combined dataset.

183

184Bayesian relaxed molecular clocks, which assume rates of molecular evolution are uncorrelated but log-
185normally distributed among lineages (Drummond et al. 2006), as implemented in BEAST v1.4.8
186(Drummond and Rambaut 2007) were used for all dating analyses. Yule and birth-death models performed
187similarly in all preliminary analyses so the birth-death model (Gernhard 2008) with a uniform prior was
188used to model cladogenesis for all final analyses. We summarized the outputs of all MrBayes and BEAST
189MCMC analyses using TRACER (version 1.4) in order to obtain parameter estimates, as well as evaluate

190 effective sample sizes (ESSs) and convergence. ESS values greater than 100 are generally regarded as being
 191 sufficient to obtain a reliable posterior distribution (Drummond et al. 2007) and we adjusted the numbers of
 192 MCMC runs to ensure that ESSs were greater than 100 for all relevant parameters in each set of analyses
 193 conducted (numbers of MCMC runs for different analyses are specified in relevant sections). ESS values
 194 typically were much larger than 100 for most parameters in each analysis. Graphical exploration of trace
 195 files for tree likelihoods and other tree-specific parameters using TRACER (version 1.4) indicated that
 196 convergence had been reached in all cases.

197

198 *Single-Fossil Cross-Validations*

199 The agreement or consistency between single fossil calibration dates and other available fossil calibrations
 200 for ten calibrated nodes (Fig. 1 – Tree Root and nodes 1 to 9) was evaluated using a modified version of the
 201 single-fossil cross-validations developed by Near et al. (2005). There were two main differences in our
 202 approach. First, rather than using fixed points for each calibration we used lognormal distributions that
 203 placed a hard minimum bound and soft maximum bound on each calibration (Table 1), thereby allowing for
 204 uncertainty in the fossil dates (Yang and Rannala 2006; Ho and Phillips 2009). For each single fossil
 205 calibration (i) we calculated the metrics \bar{D}_x , SS_x and s (Near et al. 2005) for the other nine fossil-calibrated
 206 nodes on the tree using age estimates obtained from BEAST. We conducted the cross-validations using both
 207 the mean and median age estimates in order to evaluate whether the posterior age distributions (rather than
 208 point age estimates) influenced which fossil calibrations were identified as incongruent. The difference
 209 between the molecular and fossil age at each node was calculated as $D_i = (MA_i - FA_i)$, where FA_i is the fossil
 210 age and MA_i is the mean or median molecular age estimate for node i using the candidate fossil calibration
 211 at node x . The average difference \bar{D}_x between the molecular and fossil ages across the nine other fossil
 212 calibrated nodes for the fossil calibration at node x was then calculated as

$$213 \quad \bar{D}_x = \frac{\sum_{i \neq x} D_i}{n - 1}.$$

214 The fossil age for each candidate fossil calibrated node (x) was used as a single calibration prior in the
 215 BEAST analysis and \bar{D}_x and its SE were calculated from the remaining nine candidate fossil-dated nodes.

17

216SS values were then calculated as the sum of the squared differences between the molecular (MA) and fossil
217(FA) age estimates at all other fossil-dated nodes using the formula

$$218 SS_x = \sum_{i \neq x} D_i^2 .$$

219Finally the average squared deviations, s , were calculated using the formula

$$220 s = \frac{\sum_{x=1}^n \sum_{i \neq x} D_i^2}{n(n-1)} \text{ where } n \text{ is equal to the total number of observations of } D_i \text{ (i.e. the number of fossil calibrations}$$

221remaining). For more details about the single-fossil cross validation analyses see Near et al. (2005).

222

223The second difference in our approach was that, rather than using the cross-validations to exclude specific
224fossils, we used them in a more exploratory fashion to evaluate the alternative placements of three fossils as
225calibrations for their respective stem (nodes 4, 6 and 8) and crown (nodes 5, 7 and 9) clades (Table 1). We
226also evaluated three different pairs (referred to as calibration sets) of fossil dates for two nodes, the most
227recent common ancestor (MRCA) of Caenophidia (Fig. 1 - node 2) and the MRCA of Colubroidea (Fig. 1 –
228node3), based on their previous use in other studies (Table 1). Each alternative set of fossil dates for nodes 2
229and 3 (Table 1: *Sets A, B, C*) was evaluated by conducting a separate iteration of the cross validation
230exercise (i.e., three separate iterations). In each case, the calibration set and the corresponding molecular
231dates from the single-fossil dating analyses were used to calculate \bar{D}_x , SS_x and s . The molecular and fossil
232dates for the other eight single-fossil calibrated nodes were the same for the three calibration sets.

233

234Preliminary analyses revealed that the shallower calibrations (Fig. 1, nodes 4-9) artificially inflated age
235estimates at deeper nodes to unrealistically high values. In order to stabilize estimated ages at deeper nodes
236we constrained the root using a normal prior (mean = 110 MA, 95% CI = 85-135 MA) spanning a wide
237range of plausible dates for this node (Table 1) in all single-fossil calibration analyses. BEAST runs for
238single-fossil cross-validations were conducted as follows: nDNA - 4,000,000 generations sampled every 100
239generations; mtDNA - 5,000,000 generations sampled every 100 generations; mtDNA3rdExcl – 10,000,000
240generations sampled every 100 generations.

241

18

243 The approach of Marshall (2008) involves generating an ultrametric tree that is uncalibrated with respect to
244 the fossil record and then mapping all candidate fossil calibrations onto the tree to determine which of the
245 calibrated lineages has the best temporal fossil coverage. Specifically, the method aims to identify the
246 lineage for which the oldest fossil (for that lineage) sits proportionally closest to the node of its most recent
247 common ancestor (true time of origin), and therefore has the best temporal coverage. Marshall (2008)
248 emphasizes two assumptions of the method: 1) the proportional branch lengths of the ultrametric tree are
249 accurate and 2) fossilization is random: however, the method also assumes that fossils are accurately dated
250 and assigned correctly to their respective lineages (see below for further discussion).

251

252 The first and arguably most important step in the approach of Marshall (2008) is to generate a reliable
253 ultrametric phylogeny that is uncalibrated with respect to the fossil record using an appropriate relaxed clock
254 algorithm. Given that obtaining accurate proportional branch lengths of the ultrametric tree is critical to the
255 success of this method, we generated a number of ultrametric trees using different approaches and compared
256 the results. Specifically we generated ultrametric trees for the mtDNA and nDNA datasets in BEAST by
257 constraining the tree root with a fixed value (arbitrarily set to 100). However, MCMC runs of 20,000,000
258 generations were needed to obtain ESSs > 100 for the calibrated nodes using nDNA, and convergence could
259 not be achieved for mtDNA. As such, we followed the approach of Marshall (2008) and obtained
260 ultrametric trees using r8s (Sanderson, 2003). r8s requires user-specified input trees so we used MrBayes
261 (MCMC chains of 2,000,000 generations sampling every 100 generations and all default settings) to obtain
262 optimal Bayesian phylogenies for the nDNA and mtDNA datasets using the same partitioning strategies and
263 models of evolution used for the BEAST analyses. As there is evidence that branch lengths are more
264 accurately estimated by maximum likelihood (ML) than Bayesian criteria (Schwartz and Mueller, 2010), we
265 also generated ML trees for the nDNA, mtDNA and mtDNA3rdExcl datasets in PAUP (Swofford, 2000)
266 under optimal models of sequence evolution obtained from AIC in Modeltest (Posada and Crandall, 1998).
267 We generated rooted input trees (required by r8s) by adding sequences obtained from GenBank (Appendix
268 1) for two outgroup taxa (the lizard genera *Varanus* and *Calotes*) to the datasets. The lizard taxa were
269 pruned from the optimal ML and Bayesian trees and the resulting rooted trees used to obtain ultrametric

270trees in r8s, again fixing the root age to an arbitrary value of 100. We used semi-parametric penalised
 271likelihood (Sanderson, 2002) and optimal smoothing parameters identified from the cross-validation
 272procedure in r8s as follows: MrBayes tree - smoothing parameter of 3200 with log penalty function; ML tree
 273– smoothing parameter of 3200 with additive penalty function. Given that Smith et al. (2006) demonstrated
 274that the log penalty function better estimated branch lengths than the additive penalty function for calibrated
 275ultrametric trees, we also generated an ML ultrametric tree using the log penalty function and optimal
 276smoothing parameter of 320 (note however that the sum of squares obtained from the cross validations for
 277the log penalty function were much higher than the additive penalty function, suggesting that the additive
 278penalty was more appropriate).

279

280We used the resultant ultrametric trees to calculate the empirical scaling factor (*ESF*) for each candidate
 281fossil calibration (including the three alternative fossil dates for nodes 2 and 3 and the alternative placements
 282of three fossils, Table 1) using the equation

$$283 \text{ } ESF_i = \frac{FA_i}{NTL_i}, \text{ where } FA_i \text{ is the age of the oldest fossil of the lineage and } NTL_i \text{ is the relative node to tip}$$

284length of the branch of that lineage on the ultrametric phylogeny (Marshall, 2008). The fossil with the
 285largest *ESF_i* is regarded as having the best temporal coverage; however, fossils that have been incorrectly
 286assigned and/or incorrectly dated may also have the highest *ESF* values, and these outliers need to be
 287identified. We tested for possible fossil outliers by comparing the distribution of *ESF_i* values to a uniform
 288distribution using the Kolmogorov-Smirnov test, on the assumption that *ESF_i* values for fossil outliers lie
 289outside a uniform distribution (Marshall 2008). One limitation of this approach is that it is most effective if
 290there is just one outlier (Marshall 2008, pg 732). We were testing the alternative stem and crown
 291placements of three fossils. As such, the *ESF_i* values for the crown placements (that inevitably will be larger
 292than the *ESF_i* values for their stem placements) might potentially cluster together, thereby making it
 293impossible to identify them as outliers. In order to address this issue we modified the approach of Marshall
 294(2008) to test the alternative placements of these fossils (see Results for details).

295

296*Bayesian Analyses to Evaluating Multi-Calibration Sets*

297 We used the method of Sanders and Lee (2007) to evaluate three alternative dates for two nodes with
298 controversial fossil calibrations in a Bayesian multi-calibration framework. This method compares the prior
299 and posterior distributions of the 95% HPD intervals for each candidate calibration, particularly focusing on
300 potentially controversial calibrations of interest. In our case, the single-fossil cross-validations identified
301 plausible congruent calibration sets comprising six fossil-calibrated nodes that included nodes 2 and 3, but
302 could not distinguish between the different possible ages assigned to these two nodes (Table 1 – *Sets A, B*
303 and *C*). In addition, the ESF_i values for the same six fossil calibrated nodes indicated that none were
304 outliers. However, ESF_i values cannot be used to evaluate alternative dates for the same node because the
305 oldest date will inevitably have the highest empirical coverage, even if that date is not correct. Moreover
306 ESF_i values from different ultrametric trees identified different fossils as having the highest empirical
307 coverage (see below for details). We evaluated the alternative ages for nodes 2 and 3 using three sets of
308 BEAST multi-calibration analyses that incorporated the four congruent calibrations and the *Set A, B and C*
309 node 2 and 3 calibration ages in turn. For each analysis we compared the prior and posterior distributions of
310 all six fossil-calibrated nodes, with the expectation that the node 2 and 3 calibration set most consistent with
311 the other four fossil dated nodes would return posterior distributions for all six calibrated nodes that were
312 similar to their prior constraints (Sanders and Lee 2007). We also conducted a fourth set of analyses using
313 the four congruent fossils with no constraints on nodes 2 and 3 (*Set D*) and compared the unconstrained and
314 constrained node 2 and 3 age estimates. These four sets of BEAST analyses were conducted for nDNA,
315 mtDNA and mtDNA3rdExcl datasets, using the same lognormal priors, relaxed molecular clocks, and
316 partitioned evolutionary models as the single-fossil dating analyses. MCMC runs comprised 4,000,000
317 generations for the nuclear data, and 10,000,000 generations for both mitochondrial datasets. In each case
318 MCMC runs were sampled every 100 generations.

319

320 Given that certain combinations of priors can interact to generate unexpected effective joint priors, we also
321 performed an analysis for each calibration set without data (empty alignments) to ensure that the effective
322 priors were similar to the original priors. We assessed how informative the data were by comparing the
323 effective priors with posteriors obtained using data (Drummond et al. 2006). These analyses indicated that
324 the effective priors were similar to the original priors, and the posteriors obtained from the data departed

325 from the priors (indicating informative data).

326

327 Results

328 The final nDNA alignment had 3264 characters of which 870 were variable and 421 were parsimony
 329 informative, while the mtDNA alignment had 2221 characters of which 1368 were variable and 1193 were
 330 parsimony informative, and the mtDNA3rdExcl had 1632 characters of which 884 were variable and 578
 331 were parsimony informative. All tree topologies from PAUP* ML analyses and Bayesian MCMC searches
 332 (MrBayes and BEAST) of the nuclear and mitochondrial datasets converged on a topology (Fig. 1) highly
 333 congruent with published molecular phylogenies for the the elapid taxa (Slowinski et al. 1997; Keogh 1998;
 334 Keogh et al. 1998; Lukoschek and Keogh 2006; Wuster et al. 2007; Sanders and Lee 2008; Sanders et al.
 335 2008; Kelly et al. 2009; Pyron et al. 2010). Data matrices and relevant trees have been submitted to
 336 TreeBASE (#11272). Eight of the ten candidate calibration nodes had extremely high support with $\geq 99\%$
 337 posterior probabilities (PPs) for all analyses conducted (Fig. 1). The two nodes with poor support were node
 338 5 (typically with $\sim 80\%$ PPs for mtDNA and $< 50\%$ PPs for nDNA) and node 8 (typically with $\sim 55\%$ PPs for
 339 mtDNA and $< 50\%$ PPs for mtDNA). Other nodes with PPs $> 98\%$ are also shown on the trees (Fig. 1).
 340 Saturation plots revealed an abundance of hidden substitutions in all three codon positions of the
 341 mitochondrial dataset (Fig. 2a-d), but particularly in the third codon position (Fig. 2d).

342

343 Single-Fossil Cross-Validations

344 In all cases, the results of single-fossil cross-validations using mean and median age estimates from BEAST
 345 were highly consistent so we present only the results from the mean age estimates. Nuclear DNA cross
 346 validations produced similar results for each calibration set, with \bar{D}_x values indicating that four fossils
 347 consistently produced older molecular divergence estimates for other candidate fossil-calibrated nodes,
 348 while the other six fossils produced younger divergence estimates; however, the relative magnitude of these
 349 tendencies differed between calibration sets (Fig. 3a). Specifically, the youngest fossil dates for nodes 2 and
 350 3 (*set A*) resulted in larger molecular overestimates and smaller underestimates of fossil dates than *sets B*
 351 and *C*, which returned similar mean differences (\bar{D}_x) between the fossil and molecular dates (Fig. 3a). *SS*
 352 values ranked the four node calibrations that consistently produced older molecular divergence estimates for

353 other fossil ages as the most incongruent fossils (Fig. 4a). *Set A* calibrations produced consistently larger *SS*
 354 values for all fossil calibrated nodes than *sets B* and *C* (Fig. 4a), reflecting the larger differences (\bar{D}_x)
 355 between the molecular and fossil dates using the younger *set A* calibrations (Fig. 3a). By contrast, *SS* values
 356 for *sets B* and *C* were very similar (Fig. 4a). Sequential removal of fossil calibrations from most to least
 357 divergent, as ranked by *SS* values (Fig. 4a), resulted in steep incremental declines in *s* values for the
 358 subsequent removal of nodes 7, 9, 5 and 4 for all calibration sets (Fig. 5a). At this point *s* values for *sets B*
 359 and *C* were small and subsequent removal of fossils did not markedly decrease *s* values (Fig. 5a). Starting *s*
 360 values for *set A* were much larger than for *sets B* and *C* and did not drop to low values until the fifth fossil
 361 calibration (node 2) was removed and then remained low (Fig. 5a).

362

363 Mitochondrial DNA produced a markedly different pattern of mean differences (\bar{D}_x) between the molecular
 364 and fossil dates than nuclear DNA (Fig. 3). Most notably, the four fossil calibrations (nodes 4, 5, 7, 9) that
 365 returned much older nuclear DNA values for fossil ages at other candidate calibration nodes either produced
 366 younger or only slightly older estimates of fossil ages for mtDNA (Fig. 3b) and this remained the case even
 367 when the third codon positions were removed (Fig. 3c). In addition, the tendency for nodes 6 and 8 to
 368 produce younger molecular ages for fossil dates at other nodes was more extreme for the mitochondrial than
 369 nuclear data, and this was true for both mitochondrial datasets (Figs. 3b & c). By contrast, node 1 produced
 370 older ages at other nodes for both mtDNA datasets, whereas this node produced younger dates for nuclear
 371 DNA. Given these differences it is not surprising that mitochondrial *SS* values ranked fossils differently
 372 than nuclear *SS* values (Figs. 4b & c). In addition, \bar{D}_x values for the younger *set A* calibrations (at nodes 2
 373 and 3) did not follow the same pattern as for *sets B* and *C* (Figs. 3b & c) and the mitochondrial rank-order of
 374 candidate calibrations was different for *set A* calibrations than for *sets B* and *C*, which were similar (Figs. 4b
 375 & c). *Sets B* and *C* had highest *SS* values at nodes 6 and 8; however, removing these nodes only slightly
 376 decreased *s* values, which did not decline sharply until subsequent removals of the third and fourth ranked
 377 fossils and then remained low (Figs. 5b & c). Interestingly, node 1 was the most incongruent fossil for the
 378 younger *set A* calibrations for the entire mtDNA dataset and *s* values dropped sharply when it was removed.
 379 Subsequent removal of the three next most incongruent fossils did not produce further decreases in *s*, but *s*

380 decreased with the removal of the fifth and subsequent fossils (Fig. 5b). By contrast, node 8 was the most
381 incongruent fossil for all three calibration sets for the mtDNA dataset with third codon position excluded
382 and s values did not drop sharply until the first two most incongruent nodes were excluded in each case (Fig.
383 5c).

384

385 *Fossil Coverage and Fossil Outliers*

386 The four ultrametric trees obtained from the nDNA dataset differed in their proportional branch lengths,
387 resulting in differing ESF_i values for the candidate fossil calibrations (Table 2). Nonetheless, the four
388 highest ESF_i values (in decreasing order) for the ML and MrBayes ultrametric trees were for nodes 9, 7, 5
389 and 4 (Table 2), the same nodes identified as least congruent by the cross-validation analyses. These four
390 nodes also had the highest ESF_i values for the BEAST ultrametric tree, but in different decreasing order
391 (Table 2). Lack of resolution in the ML and Bayesian nDNA trees resulted in nodes 4 and 5 forming a
392 polytomy: as such, it was not possible to evaluate the alternative placements of this fossil calibration (as the
393 ESF_i values for the stem and crown placement were identical). Moreover, issues regarding the taxonomic
394 affinities of these fossils (Table 1 and Supplementary Material A) suggest that it is not possible to accurately
395 place them on the phylogeny (despite their use to date caenophidian divergences in previous studies:
396 Guicking et al. 2006; Alfaro et al. 2008). As such, we excluded them from the outlier analysis.

397

398 Nodes 7 and 9 were the shallower crown placements of the two candidate fossil calibrations for which the
399 alternative deeper stem placements also were evaluated. Obviously the candidate fossils cannot correctly be
400 assigned to both the stem and crown nodes so, prior to testing whether the distributions of ESF_i values
401 conformed to uniform distributions, we removed the ESF_i values for the corresponding stem placements of
402 each fossil (nodes 6 and 8). The resulting distributions of ESF_i values for the BEAST and ML ultrametric
403 trees (under both the additive and log penalty functions) were strongly rejected as belonging to uniform
404 distributions (BEAST $p < 0.05$; ML trees $p < 0.005$ in both cases); however, this was not the case for the
405 MrBayes tree ($0.20 < p > 0.10$). These inconsistent results highlight the sensitivity of this approach to
406 differences in proportional branch lengths obtained from ultrametric trees obtained using different methods
407 (see below for further discussion). Given that the weight of evidence suggested that crown placement of the

408 *Naja* fossil was an outlier, we removed the ESF_i values for node 9 and reinserted the ESF_i values for the
409 corresponding stem placement of the fossil (node 8). The resulting distributions of ESF_i values for the
410 MrBayes and ML ultrametric trees also were rejected as belonging to uniform distributions, suggesting that
411 the crown placement of the putative *Laticauda* fossil at node 7 also is an outlier. However, this was not the
412 case for the BEAST ultrametric tree (Table 2). We then removed the ESF_i values for node 7 (from the ML
413 and MrBayes ESF_i distributions) and inserted the ESF_i values for the stem placement of the fossil at node 6.
414 The resulting distributions of ESF_i values were not rejected as belonging to uniform distributions. In terms
415 of the MrBayes tree, the inclusion of ESF_i values for both potential outliers (nodes 7 and 9) may have
416 resulted in the artefact mentioned by Marshall (2008), whereby the larger ESF_i values of outliers group
417 together making it impossible to distinguish the resultant distribution from a uniform distribution (thereby
418 failing to identify node 9 as an outlier). In order to explore this possibility we removed the ESF_i for node 7
419 and retained the ESF_i of the corresponding stem placement at node 6. The resulting distribution of ESF_i
420 values did not conform to a uniform distribution, supporting node 9 as an outlier. Overestimation of shorter
421 branches has recently been demonstrated for Bayesian approaches (Schwartz and Mueller 2010), and the
422 smaller difference between ESF_i values for nodes 9 and 7 for the Bayesian than ML trees may reflect
423 overestimation of short branches in the crown *Naja* clade by MrBayes.

424

425 The proportional branch-lengths and corresponding ordering of ESF_i values for the ultrametric trees
426 obtained from optimal mtDNA ML and MrBayes and the mtDNA3rdExcl ML trees were different from
427 those obtained from nDNA (Table 2). For the both mtDNA trees, the crown nodes 5, 7 and 9 still had the
428 highest ESF_i values, while for the mtDNA3rdExcl tree the node 2 *Set C* had the highest ESF_i value (Table
429 2). However, the distributions of ESF_i values conformed to uniformity for all three mitochondrial
430 ultrametric trees (ML and MrBayes), and this result was true for distributions including just one potential
431 crown node outlier (and the corresponding stem placement of the other fossil): thus, no outliers were
432 identified.

433

434 *Evaluating Multi-Calibration Sets using Bayesian Analyses*

435 There were consistent differences in the plausible sets of congruent fossil calibrations identified from the

436 cross-validations from nuclear and mitochondrial DNA, and the fossil outliers identified from nuclear but
437 not mitochondrial data based on ESF_i values. These differences are almost certainly due to the effects of
438 nucleotide saturation for mtDNA (see Discussion). As such, we conducted the multi-calibration analyses
439 using the six fossil calibrated nodes selected by the nuclear data.

440

441 Multi-calibration analyses using nuclear DNA revealed similarities and differences between the estimated
442 mean ages and 95% Highest Posterior Densities (HPD) intervals for the six calibrated nodes across
443 calibration sets *A*, *B*, *C* and *D*. The most striking similarities were for the four fossil calibrations common to
444 each calibration set (tree root and nodes 1, 6 and 8), for which the means and minimum 95% HPD intervals
445 were very similar to their respective calibration priors (<5% in all cases), while maximum 95% HPD
446 intervals invariably were smaller than the calibrations (Fig. 6). By contrast, age estimates for nodes 2 and 3
447 differed considerably between calibration sets, in part reflecting the influence of their calibration priors but
448 also reflecting inconsistencies between these priors and the other four fossil calibrations (Fig. 6). Moreover,
449 age estimates for nodes 2 and 3 tended to converge on ages estimated by set *D* (Fig. 6), in which nodes 2 and
450 3 were not constrained. This tendency was most pronounced for node 2, for which the set *A* age estimate
451 was far more similar to the set *D* estimate than to the set *A* calibration prior. Indeed, the set *A* prior and
452 posterior distributions barely overlapped (Fig. 6). Similarly, the set *B* estimated age for node 2 also was
453 closer to the set *D* estimate than to the set *B* calibration prior, with the set *B* maximum age estimate 70
454 million years younger than its calibration prior (Fig. 6). Set *C* returned a node-2 age estimate that was
455 similar to both its calibration prior and the set *D* age estimate for this node, although its minimum 95% HPD
456 interval was younger than the hard minimum bound of the prior. The node-3 age nDNA estimates were
457 more similar to their respective calibration priors, but again, posterior distributions diverged from priors
458 towards the unconstrained set *D* age estimate. The set *A* estimated mean age was slightly older than its
459 calibration prior, but posterior and prior distributions were identical, while the set *C* age estimate also was
460 identical to the mean and minimum bounds of the calibration prior (Fig. 6). The set *B* estimated mean age
461 and minimum 95% HPD were younger than the calibration prior (Fig. 6).

462

463 Mitochondrial age estimates were invariably older for the shallower nodes 3, 6 and 8 than their respective

464 calibration priors and, with one exception, also for the corresponding nDNA age estimates. By contrast,
465 mitochondrial node-1 age estimates for all calibration sets were similar to the calibration prior and to nuclear
466 DNA age estimates, and this was true for both mitochondrial datasets (Fig. 6). Nonetheless, the tendency
467 for mtDNA to return older age estimates at shallow nodes and the tree root was much pronounced when the
468 third codon positions were excluded, with mtDNA3rdExcl age estimates for nodes 6 and 8 age intermediate
469 to the nDNA and mtDNA age estimates, and tending to converge on mean nDNA age estimates for node 3
470 and the tree root (Fig. 6). While mitochondrial age estimates for node 2 from the entire dataset showed the
471 same tendency as nuclear ages to converge on the unconstrained *set D* age estimates (irrespective of the
472 calibration prior used), this was not the case for mtDNA with third codon positions excluded (Fig. 6).
473 Indeed, with the exception of *set C*, the node-2 mtDNA3rdExcl age estimates tended to converge on the
474 calibration prior resulting in age estimates that were younger than the corresponding nDNA estimates, and
475 this was also true for the node-3 *set A* age estimate (Fig. 6).

476

477 Discussion

478 Increasing awareness of the importance of identifying reliable fossils to calibrate molecular clocks has
479 resulted in the development of several methods for evaluating and employing fossil calibrations (reviewed
480 by Ho and Phillips 2009). Each approach has advantages and limitations, as we demonstrate by comparing
481 three different approaches with particular emphasis on the impact of nucleotide saturation on the fossils
482 selected.

483

484 The cross-validation method (Near et al. 2005) discards calibrations until an internally consistent set is
485 obtained, and in the process, may discard calibrations with the best temporal coverage because they are
486 inconsistent with the remaining calibrations. Nonetheless, the method has been used in several recent
487 studies (Near and Sanderson 2004; Noonan and Chippendale 2006b; Rutschmann et al. 2007; Alfaro et al.
488 2008). By contrast, the use of empirical scaling factors aims to identify one fossil with the best empirical
489 coverage (Marshall 2008); however, accurate results are highly dependant on meeting the assumptions of the
490 method (see below). Unlike the cross-validation approach, empirical scaling factors (*ESFs*) have only been
491 used in one previous study (Davis et al. 2009). This study obtained an ultrametric tree in r8s using penalised

492likelihood with log penalty function (following the advice of Marshall 2008), based on empirical evidence
493that penalised likelihood (PL) using the log penalty function produces the most reliable ultrametric trees
494(Smith et al. 2006). However, Davies et al. (2009) comment that their resultant dates were much older than
495expected for several lineages. Our study demonstrated that ultrametric trees generated from ML and
496Bayesian nDNA phylogenies using the log penalty function were incongruent in terms of the magnitude and
497order of the *ESFs* (Table 2) and the fossil outliers identified. By contrast, results from the ML ultrametric
498tree using the additive penalty function were more similar to those obtained for the MrBayes tree. At the
499very least, these results suggest that the findings of Smith et al. (2006) are not universal and various
500approaches for obtaining uncalibrated ultrametric trees need to be evaluated for reliability and consistency of
501results.

502

503These conflicting results highlight a major limitation of *ESFs*, which is the reliance on accurate proportional
504branch lengths (which we do not know, or the entire dating process would be considerably easier). The final
505step of Marshall's (2008) approach uses the lineage with the highest coverage to calibrate the tree and
506estimate divergences. Our nuclear DNA results suggest that the *set B* date for node 3 had the highest
507coverage (Table 2). However, we were evaluating several controversial fossil ages for this node (Table 1)
508and, by default, the highest coverage will be assigned to the oldest fossil so *ESFs* cannot be used for this
509task.

510

511The third method we evaluated, which uses a Bayesian framework to evaluate several candidate fossils in a
512multi-calibration framework (Sanders and Lee 2007), is ideally suited for the task. However, one limitation
513of this method is that at least some of the candidate calibrations are assumed to be reliable, with just one or
514two calibrations being evaluated. In addition, multiple calibrations can interact with each other to generate
515different effective priors; however, the extent of this effect can be evaluated explicitly (Drummond et al.
5162006) and our analyses of priors with empty alignments indicated that this was not an issue in our study.
517Nonetheless, one limitation of our study was that the calibrations for nodes 2 and 3 were evaluated in pairs
518based on their previous use in other studies and, as such, the best combination may not have been included
519in our analyses. Rutschmann et al., (2007) recently presented an alternative approach for evaluating the

520 internal consistency of fossil calibrations that compared s values from all possible combinations of dates and
521 nodes (72 combinations in our case) (Rutschmann et al. 2007). However, this approach will be subject to
522 the same saturation effects demonstrated in our study and, as such, the effects of using rapidly and slowly
523 evolving gene regions or codon positions for evaluating the internal consistency of calibrations will need to
524 be considered.

525

526 There is a growing consensus that the advantages of using multiple independent fossil calibrations
527 significantly outweigh any disadvantages (Ho and Phillips 2009). Multiple calibrations can ameliorate the
528 effects of errors in fossil dates and/or the assignment of fossils to certain nodes (Conroy and van Tuinen
529 2003; van Tuinen and Dyke 2004), provided that errors are not biased in the same direction. Moreover, the
530 use of multiple calibrations allows the explicit modelling of rate variation among lineages. The limitations
531 of using just one calibration in BEAST analyses for modelling rate variation are highlighted in the
532 chronogram from the mitochondrial dataset with third codon positions removed: the two basal branches
533 extending from the tree root on the BEAST chronogram were massively stretched, and the remaining
534 internal branches overly compressed (Fig. 1c). The addition of multiple calibrations ameliorated this effect
535 (Fig. 6), presumably resulting in more accurately estimated branch lengths (time) throughout the
536 chronogram. Although the mtDNA3rdExcl ultrametric tree generated in r8s did not suffer from similarly
537 stretched basal branches (results not shown), the approach of Marshall (2008) ultimately relies on just one
538 calibration to date the phylogeny and our analyses demonstrated the highly variable results that could be
539 obtained using different methods to generate the ultrametric tree (Table 2). Moreover, while this approach
540 might be realistic for groups with exceptionally good fossil records (provided that the hurdle of obtaining a
541 reliable ultrametric tree can be overcome), on its own it is likely to produce highly misleading results in the
542 majority of cases where the fossil record is less than ideal.

543

544 *Evaluating the Effects of Saturation on Identifying Reliable Calibrations*

545 The differences in the plausible sets of congruent fossil calibrations identified from the cross-validations
546 from nuclear and mitochondrial DNA, as well as fossil outliers identified from nuclear but not mitochondrial
547 data based on ESF_i values, can be entirely accounted for by saturation effects. The saturation plots revealed

548strong mitochondrial saturation in the dataset (Fig. 2), particularly the third codon position (Fig. 2d). The
549saturation effects on tree topology, and corresponding age estimates of fossil calibrated nodes, are clearly
550evident in Figure 1. Compared with the nuclear chronogram (Fig. 1a), the chronogram from the entire
551mitochondrial dataset had compressed internal branches, which essentially reduced the total distance (time)
552between nodes 1 and 9 on the chronogram (Fig. 1b). This result was also true for the nDNA and mtDNA
553ultrametric trees generated in r8s (not shown).

554

555In terms of the cross-validations, the three sets of nuclear cross-validations identified the same four shallow
556fossil calibrated nodes (4, 5, 7 and 9) as least congruent with the six other candidate calibrations tested.
557These nodes also had the highest ESF_i values (Table 2), with nodes 7 and 9 being identified as outliers by
558three of the four nuclear DNA ultrametric trees. By contrast, mitochondrial cross validations identified
559nodes 6 and 8 as least congruent for *sets B* and *C* (and also *set A* when the third codon positions were
560removed). Thus, for two fossils (*Naja* and *Laticauda*) nuclear DNA favored stem placement (nodes 8 and 8)
561while mtDNA favored crown placement (nodes 7 and 9), directly as the result of saturation effects.
562Specifically, if a crown group is constrained with the same fossil calibration as its respective stem group, the
563placement of a fossil at the shallower crown node will return older estimates at other nodes than stem
564placement, irrespective of data type. However, because mitochondrial distances were artificially shortened
565(due to compression of internal branches resulting from nucleotide saturation) the tendency for crown
566placement to produce much older age estimates for other fossil calibrated nodes, which was so strongly
567apparent for nuclear DNA, disappeared for mtDNA: instead, stem placement resulted in younger age
568estimates at deeper fossil calibrated nodes. Similarly, the compressed internal branches for mtDNA resulted
569in smaller differences between the larger ESF_i values; thus ESF_i distributions did not deviate from
570uniformity with the result that fossil outliers were not identified. Evaluating these results in terms of the
571actual fossils (Table 1 and Supplementary Material A) further suggests that misleading results were obtained
572from the mitochondrial data due to the effects of saturation.

573

574The effects of mitochondrial saturation are also evident in many studies estimating divergence times in
575snakes. Studies that have relied primarily or entirely on mitochondrial data (Nagy et al. 2003; Guicking et

576al. 2006; Burbrink and Lawson 2007; Wuster et al. 2007, 2008; Alfaro et al. 2008; Kelly et al. 2009), have
577recovered two-fold older age estimates for some advanced snake clades from mitochondrial sequence data
578(see Table 1 in Kelly et al. 2009) than from nuclear sequence data (Sanders and Lee 2008), even when
579almost exactly the same calibrations were used (Sanders and Lee 2008; Kelly et al. 2009). Jiang et al.
580(2007) demonstrated accelerated rates of mitochondrial evolution in advanced snakes, suggesting that the
581extent of nucleotide saturation may be more pronounced than in other taxonomic groups. Nonetheless, the
582effects of mitochondrial saturation for estimating branch lengths and dating divergences have been well
583documented for other vertebrate groups such as agamid lizards (Hugall and Lee 2004); squamates
584(Townsend et al. 2004); tetrapods (Hugall et al. 2007); rodents (Jansa et al. 2006); and across all vertebrates
585(Phillips 2009). In addition, Brandley et al. (2011) recently demonstrated the importance of data partitioning
586for obtaining accurate divergence estimates in lizards, particularly drawing attention to the effects of highly
587saturated mitochondrial third codon positions. We found similarly high levels of third codon mitochondrial
588saturation (Fig. 2d), yet removing third codon nucleotides did not ameliorate saturation effects for the
589single-fossil cross validations (Figs. 2-4). However, removing third codon nucleotides improved the multi-
590calibration BEAST analyses. Posterior distributions for the six fossil-calibrated nodes (with third codon
591nucleotides excluded) typically converged on age estimates from nuclear DNA (deeper nodes) or were
592intermediate between the mitochondrial (entire) and nuclear DNA results (shallower nodes). These results
593suggest that removing highly saturated third codon positions in multi-calibration Bayesian analyses might
594provide a way forward for dealing with mitochondrial saturation, both for evaluating fossil calibrations and
595for estimating divergences. More importantly our study demonstrates that saturation strongly influenced
596different approaches for evaluating candidate calibrations and highlights the need to carefully consider the
597effects of data type when evaluating fossils.

598

599*How Wrong Can We Be?*

600Pulquerio and Nichols (2006) explored the many factors that can contribute to the highly variable dates
601obtained using calibrated molecular clocks and posed the question ‘*how wrong can we be?*’ Given that the
602choice of fossil calibrations is fundamental obtaining accurate dates, it is vital that the methods used to
603evaluate candidate fossils are used with a clear understanding of the advantages and disadvantages of each,

604as well as the effects of data type and other factors on the results. Our study highlighted the fact that
605nucleotide saturation strongly influences which fossil calibrations are identified as outliers by the cross-
606validations and empirical scaling factors. Previous studies that used cross-validations to evaluate fossil
607calibrations have tended to use some combination of nuclear and mitochondrial DNA (Near and Sanderson
6082004; Near et al. 2005; Noonan and Chippindale 2006b; Alfaro et al. 2008) or nuclear and plastid DNA
609(Rutschmann et al. 2007), and this was also the case for the fossil coverage approach (Davies et al. 2009;
610Marshall 2008). We also conducted many of the Bayesian single and multi-calibration analyses using a
611combined mitochondrial and nuclear dataset (with appropriate partitioning), and the results were very
612similar to those of the mitochondrial data (results not shown), indicating that combining nuclear and
613mitochondrial data does not inevitably counteract the effects of mitochondrial saturation (but see Brandley et
614al. 2011). Given that nucleotide saturation typically has the effect of compressing basal branches, it is most
615likely that older calibrations at shallow nodes will be identified as more congruent with candidate
616calibrations at deeper nodes by cross-validations using sequence data with high levels of saturation, yet not
617be identified as outliers based on the distribution of empirical scaling factors, as was the case in our study.
618If these calibrations subsequently are used in dating analyses that also rely partially or entirely on saturated
619DNA, the resultant age estimates will suffer from the compounded effects of two sources of error from
620nucleotide saturation. Recent studies have demonstrated the potential benefits of appropriate data
621partitioning (Brandley et al. 2011) and the use of RY coding for mitochondrial data (Phillips, 2009) for
622ameliorating saturation effects on estimating divergence dates. We demonstrate that excluding third codon
623positions can also ameliorate saturation effects in Bayesian multi-calibration analyses, with relevance both
624for evaluating fossil calibrations and estimating divergences.

625

626To our knowledge there has been no previous evaluation of the effects of data type (saturation) on
627approaches for evaluating fossil calibrations (Near and Sanderson 2004; Rutschmann et al. 2007; Sanders
628and Lee 2007). Given that these approaches are in their infancy, further exploration of the effects of using
629sequence data with different evolutionary rates for evaluating candidate fossils and their most appropriate
630placement on a phylogeny is obviously needed. In the meantime, we urge researchers evaluating candidate
631fossil calibrations to utilise several of the methods currently available and critically compare the results.

632 Moreover, we think it imperative that researches conduct these analyses using separate nuclear and
 633 mitochondrial datasets (rather than combining the data) and use one or more of the various approaches for
 634 ameliorating mitochondrial saturation and compare the results, particularly when evaluating fossils that span
 635 very different temporal depths on the tree.

636

637 Acknowledgements

638 This study was supported by funds from the University of California, Irvine. We thank Matt Phillips for
 639 insightful comments and for invaluable assistance in generating ultrametric trees in r8s. We thank Andrei
 640 Tatarenkov, Jim McGuire, Frank Anderson and one anonymous reviewer for extensive constructive
 641 comments that significantly improved the manuscript. JSK thanks the Australian Research Council for
 642 ongoing support. VL thanks the ARC Centre of Excellence for Coral Reef Studies at JCU for support while
 643 finalizing the manuscript.

644

645 References

- 646 Albino, A. M. 2000. New record of snakes from the Cretaceous of Patagonia (Argentina). *Geodivers.*
 647 22:247-253.
- 648 Alfaro, M.E., D.R. Karns, H.H. Voris, C.D. Brock, B.L. Stuart. 2008. Phylogeny, evolutionary history, and
 649 biogeography of Oriental-Australian rear-fanged water snakes (Colubroidea: Homalopsidae) inferred
 650 from mitochondrial and nuclear DNA sequences. *Mol. Phylogenet. Evol.* 46:576-593.
- 651 Benton, M.J., F.J. Ayala. 2003. Dating the tree of life. *Science* 300:1698-1700.
- 652 Brandley, M. C., Y. Wang, X. Guo, A. N. Montes de Oca, M. Feria-Ortiz, T. Hikida, H. Ota. 2001.
 653 Accommodating heterogeneous rates of evolution in molecular divergence dating methods: an
 654 example using intercontinental dispersal of *Plestiodon (Eumeces)* lizards. *Syst Biol* 60:3-15.
- 655 Burbrink, F.R., R. Lawson. 2007. How and when did Old World ratsnakes disperse into the New World?
 656 *Mol. Phylogenet. Evol.* 43:173-189.
- 657 Conroy, C.J., M. van Tuinen. 2003. Extracting time from phylogenies: positive interplay between fossil and
 658 genetic data. *J. Mammal.* 84:444-455.
- 659 Davis, R. B., S. L. Baldauf, P. J. Mayhew. 2009. Eusociality and the success of the termites: insights from a

- 49
660 supertree of dictyopteran families. *J. Evol. Biol.* 22:1750-1761.
- 661 Donoghue, M.J., M.J. Benton. 2007. Rocks and clocks: calibrating the Tree of Life using fossils and
662 molecules. *Trends Ecol. Evol.* 22:424-431.
- 663 Doyle, J.A., M.J. Donoghue. 1993. Phylogenies and angiosperm diversification. *Paleobiology* 19:141-167.
- 664 Drummond, A.J., S.Y.W. Ho, M.J. Phillips, A. Rambaut. 2006. Relaxed phylogenetics and dating with
665 confidence. *PLoS Biol.* 4:699-710.
- 666 Drummond, A.J., A. Rambaut. 2007. BEAST: Bayesian evolutionary analysis by sampling trees. *BMC*
667 *Evol. Biol.* 7:214.
- 668 Drummond, A.J., S.Y.W. Ho, N. Rawlence, A. Rambaut. 2007. A Rough Guide to BEAST 1.4. 1-42.
- 669 Forstner, M.R.J., S.K. Davis, E. Arevalo. 1995. Support for the hypothesis of anguimorph ancestry for the
670 suborder serpentes from phylogenetic analysis of mitochondrial DNA sequences. *Mol. Phylogenet.*
671 *Evol.* 4:93-102.
- 672 Gardner, J. D., R. L. Cifelli. 1999. A primitive snake from the Cretaceous of Utah. *Special Papers*
673 *Palaeontol.* 60:87-100.
- 674 Gernhard, T. 2008. The conditioned reconstructed response. *J. Theor. Biol.* 253:769-778.
- 675 Graur, D., W. Martin. 2004. Reading the entrails of chickens: molecular timescales of evolution and the
676 illusion of precision. *Trends Genet.* 20:80-86.
- 677 Groth, J.G., G.F. Barrowclough. 1999. Basal divergences in birds and the phylogenetic utility of the nuclear
678 RAG-1 gene. *Mol. Phylogenet. Evol.* 12:115-123.
- 679 Guicking, D., R. Lawson, U. Joger, M. Wink. 2006. Evolution and phylogeny of the genus *Natrix*
680 (Serpentes: Colubridae). *Biol. J. Linn. Soc.* 87:127-143.
- 681 Head, J., P.A. Holroyd, J.H. Hutchison, R.L. Ciochon. 2005. First report of snakes (Serpentes) from the late
682 middle Eocene Pondaung formation, Myanmar. *J. Vertebr. Paleontol.* 25:246-250.
- 683 Ho, S.Y.W., M.J. Phillips. 2009. Accounting for calibration uncertainty in phylogenetic estimation of
684 evolutionary divergence times. *Syst. Biol.* 58:367-380.
- 685 Holman, J.A. 1984. *Texasophis galbreathi*, new species, the earliest New World colubrid snake. *J. Vertebr.*
686 *Paleontol.* 3:223-225.
- 687 Hugall, A.F., R. Foster, M.S.Y. Lee. 2007. Calibration choice, rate smoothing, and the pattern of tetrapod

51
688 diversification according to the long nuclear gene RAG-1. *Syst. Biol.* 56:543-563.

689 Hugall, A.F., M.S.Y. Lee. 2004. Molecular claims of Gondwanan age for Australian agamid lizards are
690 untenable. *Mol. Biol. Evol.* 21:2102-2110.

691 Hugall, A.F., M.Y.L. Lee. 2007. The likelihood node density effect and consequences for evolutionary
692 studies of molecular rates. *Evolution* 61:2293-2307.

693 Jansa, S.A., F.K. Barker, L.R. Heaney. 2006. The pattern and timing of diversification of Philippine endemic
694 rodents: evidence from mitochondrial and nuclear gene sequences. *Syst. Biol.* 55: 73-88.

695 Kelly, C.M.R., N.P. Barker, M.H. Villet, D.G. Broadley. 2009. Phylogeny, biogeography and classification
696 of the superfamily Elapoidea: a rapid radiation in the late Eocene. *Cladistics* 25:38-63.

697 Keogh, J.S. 1998. Molecular phylogeny of elapid snakes and a consideration of their biogeographic history.
698 *Biol. J. Linn. Soc.* 63:177-203.

699 Keogh, J.S., R. Shine, S. Donnellan, 1998. Phylogenetic relationships of terrestrial Australo-Papuan elapid
700 snakes (Subfamily Hydrophiinae) based on cytochrome *b* and 16S r RNA sequences. *Mol.*
701 *Phylogenet. Evol.* 10: 67-81.

702 Kuch, U., J. Muller, C. Modden, D. Mebs. 2006. Snake fangs from the lower Miocene of Germany;
703 evolutionary stability of perfect weapons. *Naturewissenschaften* 93:84-87.

704 Labarga, A., F. Valentin, M. Andersson, R. Lopez. 2007. Web Services at the European Bioinformatics
705 Institute. *Nucleic Acids Res. Web Services Issue* 1-6.

706 Lawson, R., J.B. Slowinski, B.I. Crother, F.R. Burbrink. 2005. Phylogeny of the Colubroidea (Serpentes):
707 New evidence from mitochondrial and nuclear genes. *Mol. Phylogenet. Evol.* 37:581-601.

708 Lee, M.S.Y. 1999. Molecular clock calibrations and metazoan divergence dates. *J. Mol. Evol.* 49:385-391.

709 Lukoschek, V., J.S. Keogh. 2006. Molecular phylogeny of sea snakes reveals a rapidly diverged adaptive
710 radiation. *Biol. J. Linn. Soc.* 89:523-539.

711 Lukoschek, V., M. Waycott, H. Marsh. 2007. Phylogeographic structure of the olive sea snake, *Aipysurus*
712 *laevis* (Hydrophiinae) indicates recent Pleistocene range expansion but low contemporary gene flow.
713 *Mol. Ecol.* 16:3406-3422.

714 Magallon, S., M.J. Sanderson. 2001. Absolute diversification rates in angiosperm clades. *Evolution*
715 55:1762-1780.

- 716 Marshall, C.R. 2008. A simple method for bracketing absolute divergence times on molecular phylogenies
717 using multiple fossil calibration points. *Am. Nat.* 171:726-742.
- 718 McDowell, J.R. 1987. Systematics. In: *Snakes - ecology and evolutionary biology* (eds. R.A. Seigel, J.T.
719 Collins, S.S. Novak), pp. 3-50. New York: Macmillan.
- 720 McGuire, J. A., C. C. Witt, D. L. Altshuler, J. V. Remsen Jr. 2007. Phylogenetic systematics and
721 biogeography of hummingbirds: Bayesian and maximum likelihood analyses of partitioned data and
722 selection of an appropriate partitioning strategy. *Syst. Biol.* 56:837-856.
- 723 Nagy, Z.T., U. Joger, M. Wink, F. Glaw, M. Vences. 2003. Multiple colonisation of Madagascar and
724 Socotra by colubrid snakes: evidence from nuclear and mitochondrial gene phylogenies. *Proc. R.*
725 *Soc. Lond. B. Biol. Sci.* 270:2613-2621.
- 726 Near, T.J., P.A. Meylan, H.B. Shaffer. 2005. Assessing concordance of fossil calibration points in
727 molecular clock studies: an example using turtles. *Am. Nat.* 165:137-146.
- 728 Near, T.J., M.J. Sanderson. 2004. Assessing the quality of molecular divergence time estimates by fossil
729 calibrations and fossil-based model selection. *Philos. Trans. R. Soc. Lond. B. Biol. Sci.* 359:1477-
730 1483.
- 731 Noonan, B.P., P.T. Chippindale. 2006a. Dispersal and vicariance: the complex evolutionary history of boid
732 snakes. *Mol. Phylogenet. Evol.* 40:347-358.
- 733 Noonan, B.P., P.T. Chippindale. 2006b. Vicariant origin of Malagasy reptiles supports late Cretaceous
734 Antarctic land bridge. *Am. Nat.* 168:730-741.
- 735 Palumbi, S.R. (1996) *Nucleic Acids II: The Polymerase Chain Reaction*. In: *Molecular Systematics* (eds.
736 D.M. Hillis, C. Moritz, M. B.K.), pp. 205-247. Sunderland, MA: Sinauer Associates, Inc.
- 737 Parmley, D., J.A. Holman. 2003. *Nebraskophis* HOLMAN from the Late Eocene of Georgia (USA), the
738 oldest known North American colubrid snake. *Acta Zool. Cracov.* 46:1-8.
- 739 Phillips, M.J. 2009. Branch-length estimation bias misleads molecular dating for a vertebrate mitochondrial
740 phylogeny. *Gene* 441:132-140.
- 741 Posada, D., K.A. Crandall. 1998. MODELTEST: testing the model of DNA substitutions. *Bioinformatics*
742 14:817-818.
- 743 Pulquerio, M.J.F., R.A. Nichols. 2006. Dates from the molecular clock: how wrong can we be? *Trends*

- 55
744 Ecol. Evol. 22:180-184.
- 745Pyron R.A., F.T. Burbrink, G.R. Colli, A.N. Montes de Oca, L.J. Vitt, C.A. Kuczynski, J.J. Wiens. (2010)
746 The phylogeny of advanced snakes (Colubroidea), with discovery of a new subfamily and
747 comparison of support methods for likelihood trees. *Mol. Phylogenet. Evol.*
748 doi:10.1016/j.ympev.2010.11.006
- 749Rage, J.C. (1984) *Serpentes*. In: *Encyclopedia of Paleoherpetology*, pp. 1-79. Stuttgart: Gustav Fischer
750 Verlag.
- 751Rage, J.C. (1987) Fossil history. In: *Snakes - ecology and evolutionary biology* (eds. R.A. Seigel, J.T.
752 Collins, S.S. Novak), pp. 51-76. New York: Macmillan.
- 753Rage, J.C. 1988. The oldest known colubrid snakes. State of the art. *Acta Zool. Cracov.* 31:457-474.
- 754Rage, J.C. 2001. Fossil snakes from the Paleocene of Sao Jose de Itaborai, Brazil. II Boidae. *Palaeover.*
755 30:111-150.
- 756Rage, J.C., E. Buffetaut, H. Buffetaut-Tong, Y. Chaimanee, S. Ducrocq, J.J. Jaeger, V. Suteethorn. 1992. A
757 colubrid in the late Eocene of Thailand: The oldest known Colubridae (Reptilia, Serpentes). *C. R.*
758 *Acad. Sci.* 314:1085-1089.
- 759Rage, J.C., S.S.Gupta, G.V.R. Prasad. 2001. Amphibians and squamates from the Neogene Siwalik beds of
760 Jammu and Kashmir, India. *Palaeontologische Zeitschrift* 75:197-208.
- 761Rage, J.C., C. Werner. 1999. Mid-Cretaceous (Cenomanian) snakes from Wadi Abu Hashim, Sudan: The
762 earliest snake assemblage. *Palaeontol. Afr.* 35:85-110.
- 763Rambaut, A., L. Bromham. 1998. Estimating divergence times from molecular sequences. *Mol. Biol. Evol.*
764 15:442-448.
- 765Ronquist, F., J.P. Huelsenbeck. 2003. MRBAYES 3: Bayesian phylogenetic inference under mixed models.
766 *Bioinformatics* 19:1572-1574.
- 767Rutschmann, F., T. Eriksson, K.A. Salim, E. Conti. 2007. Assessing calibration uncertainty in molecular
768 dating: the assignment of fossils to alternative calibration points. *Syst. Biol.* 56:591-608.
- 769Saint, K.M., C.C. Austin, S. Donnellan, M.N. Hutchison. 1998. C-mos, a nuclear marker useful for
770 squamate phylogenetic analysis. *Mol. Phylogenet. Evol.* 10:259-263.
- 771Sanders, K., M.S.Y. Lee. 2007. Evaluating molecular clock calibrations using Bayesian analyses with soft

- 57
772 and hard bounds. *Biol. Lett.* 3:275-279.
- 773 Sanders, K.L., M.Y.L. Lee. 2008. Molecular evidence for a rapid late-Miocene radiation of Australasian
774 venomous snakes (Elapidae: Colubroidea). *Mol. Phylogenet. Evol.* 46:1180-1188.
- 775 Sanders, K.L., M.Y.L. Lee, R. Leijds, R. Foster, J.S. Keogh. 2008. Molecular phylogeny and divergence
776 dates for Australasian elapids and sea snakes (Hydrophiinae): Evidence from seven genes for rapid
777 evolutionary radiations. *J. Evol. Biol.* 21:682-695.
- 778 Sanderson, M.J. 1997. A nonparametric approach to estimating divergence times in the absence of rate
779 constancy. *Mol. Biol. Evol.* 14:1218-1231.
- 780 Sanderson, M. J. 2002. Estimating absolute rates of molecular evolution and divergence times: A penalized
781 likelihood approach. *Mol. Biol. Evol.* 19:101-109.
- 782 Sanderson, M. J. 2003. r8s: inferring absolute rates of molecular evolution and divergence times in the
783 absence of a molecular clock. *Bioinformatics* 19:301-302.
- 784 Scanlon, J.D., M.S.Y. Lee, M. Archer. 2003. Mid-Tertiary elapid snakes (Squamata, Colubroidea) from
785 Riversleigh, northern Australia: early steps in a continent-wide adaptive radiation. *Geobios* 36:573-
786 601.
- 787 Slowinski, J.B., J.S. Keogh. 2000. Phylogenetic relationships of elapid snakes based on cytochrome *b*
788 mtDNA sequences. *Mol. Phylogenet. Evol.* 15:157-164.
- 789 Slowinski, J.B., A. Knight, A.P. Rooney. 1997. Inferring species trees from gene trees: A phylogenetic
790 analysis of the Elapidae (Serpentes) based on the amino acid sequences of venom proteins. *Mol.*
791 *Phylogenet. Evol.* 8:349-362.
- 792 Smith, A. B., D. Pisani, J. A. Mackenzie-Dodds, B. Stockley, B. L. Webster, D. T. J. Littlewood. 2006.
793 Testing the molecular clock: molecular and paleontological estimates of divergence times in the
794 Echinoidea (Echinodermata). *Mol. Biol. Evol.* 23:1832-1851.
- 795 Swofford, D. 2000. PAUP*. Phylogenetic Analysis Using Parsimony (*and other methods). 4.0 ed.
796 Sunderland, MA: Sinauer Associates, Inc.
- 797 Szyndlar, Z., J.C. Rage. 1990. West Palearctic cobras of the genus *Naja* (Serpentes: Elapidae):
798 interrelationships among extinct and extant species. *Amphib-Reptilia* 11:385-400.
- 799 Szyndlar, Z., J.C. Rage. 1999. Oldest fossil vipers (Serpentes: Viperidae) from the Old World. *Kaupia* 8:9-

- 801 Townsend, T.M., A. Larson, E. Louis, J.R. Macey. 2004. Molecular phylogenetics of squamata: the position
802 of snakes, amphisbaenians, and dibamids, and the root of the squamate tree. *Syst. Biol.* 53:735-757.
- 803 van Tuinen, M., G.J. Dyke. 2004. Calibration of galliform molecular clocks using multiple fossils and
804 genetic partitions. *Mol. Phylogenet. Evol.* 30:74-86.
- 805 Vidal, N., A.S. Delmas, P. David, C. Cruaud, A. Couloux, S.B. Hedges. 2007. The phylogeny and
806 classification of caenophidian snakes inferred from seven nuclear protein-coding genes. *C. R. Biol.*
807 330:182-187.
- 808 Wuster, W., S. Crookes, I. Ineich, Y. Mane, C.E. Pook, J.-F. Trape, D.G. Broadley. 2007. The phylogeny of
809 cobras inferred from mitochondrial DNA sequences: evolution of venom spitting and the
810 phylogeography of the African spitting cobras (Serpentes: Elapidae: *Naja nigricollis* complex). *Mol.*
811 *Phylogenet. Evol.* 45:437-453.
- 812 Wuster, W., L. Peppin, C.E. Pook, D.E. Walker. 2008. A nesting of vipers: phylogeny and historical
813 biogeography of the Viperidae (Squamata: Serpentes). *Mol. Phylogenet. Evol.* 49:445-459.
- 814 Yan, J., H. Li, K. Zhou. 2008. Evolution of the mitochondrial genome in snakes: gene rearrangements and
815 phylogenetic relationships. *BMC Genet.* 9:569.
- 816 Yang, Z., B. Rannala. 2006. Bayesian estimation of species divergence times under a molecular clock using
817 multiple fossil calibrations with soft bounds. *Mol. Biol. Evol.* 23:212-226.

818 **Figure Legends**

819 **Figure 1:** Bayesian chronograms from BEAST analyses with tree roots constrained to 97 (92-120) Myr
 820 indicating the position of ten candidate fossil calibrated nodes (root and nodes 1-9) evaluated in this study.
 821 Solid black dots indicate nodes with $\geq 98\%$ posterior probabilities. *Fig. 1a* Chronogram from nuclear DNA.
 822 *Fig. 1b* Chronogram from entire mitochondrial dataset; and *Fig. 1c* Chronogram from mitochondrial data
 823 with third codon positions removed.

824

825 **Figure 2:** Saturation plots of genetic distances corrected for multiple substitutions versus uncorrected 'p'
 826 distances. Corrected genetic distances were calculated using the estimated best-fit models of sequence
 827 evolution obtained from AIC criterion in ModelTest. *Fig. 2a:* Saturation plots of the entire mitochondrial
 828 DNA dataset (black circles) versus nuclear DNA (gray diamonds). Note the different axis scales for the
 829 nuclear and mitochondrial datasets. Saturation plots are also shown for *b)* mtDNA first codon position, *c)*
 830 mtDNA second codon position, and *d)* mtDNA third codon position for the combined ND4 and cytochrome
 831 *b* genes. Note the different X-axis scales for b, c and d.

832

833 **Figure 3:** Histogram of the mean differences \bar{D}_x and standard errors (SE) between fossil and estimated
 834 molecular ages (Myr) for each of three sets of ten single-fossil calibrated nodes from *a)* nuclear DNA; *b)*
 835 mitochondrial DNA; and *c)* mitochondrial DNA with third codon position removed. Fossil ages for eight of
 836 the ten candidate nodes were identical for each set, differing only for nodes 2 and 3 (see Fig. 1). Fossil ages
 837 used as constraints are given in Table 1. For a single node (*x*) the fossil age at node *x* was used as a single
 838 calibration prior. Molecular age estimates were obtained for the nine other candidate nodes for which fossil
 839 ages were available.

840

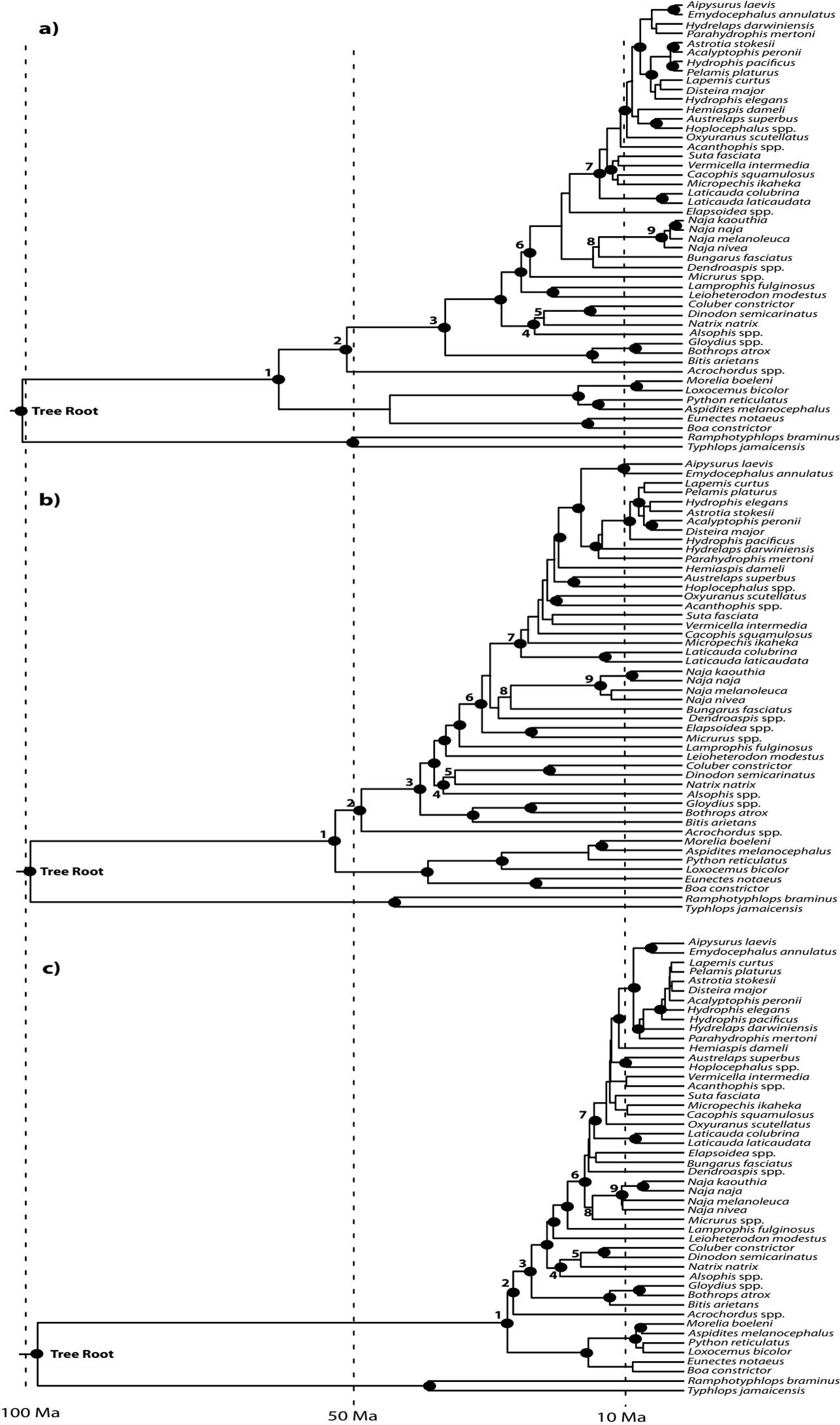
841 **Figure 4:** SS values for each candidate fossil calibration node when used as the single calibration prior in
 842 each of the three calibration sets for *a)* nuclear DNA; *b)* mitochondrial DNA; and *c)* mitochondrial DNA
 843 with third codon position removed.

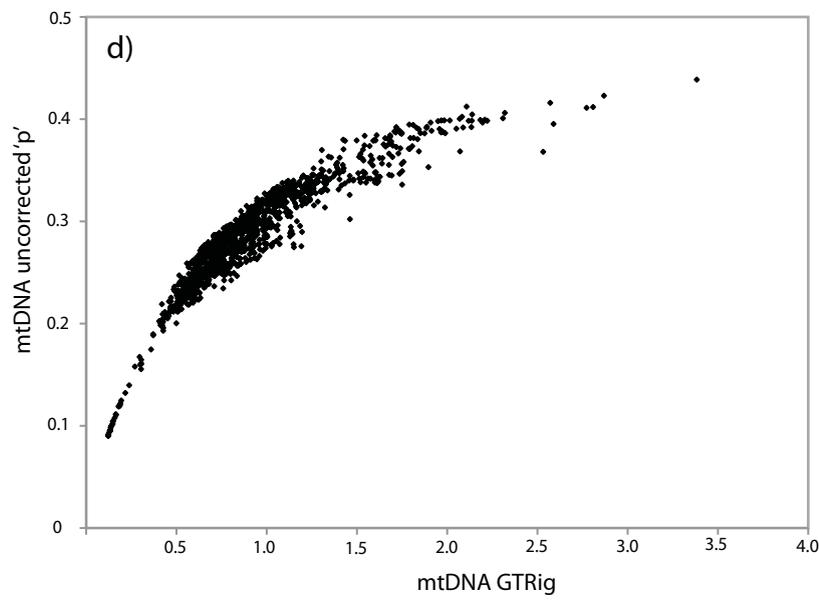
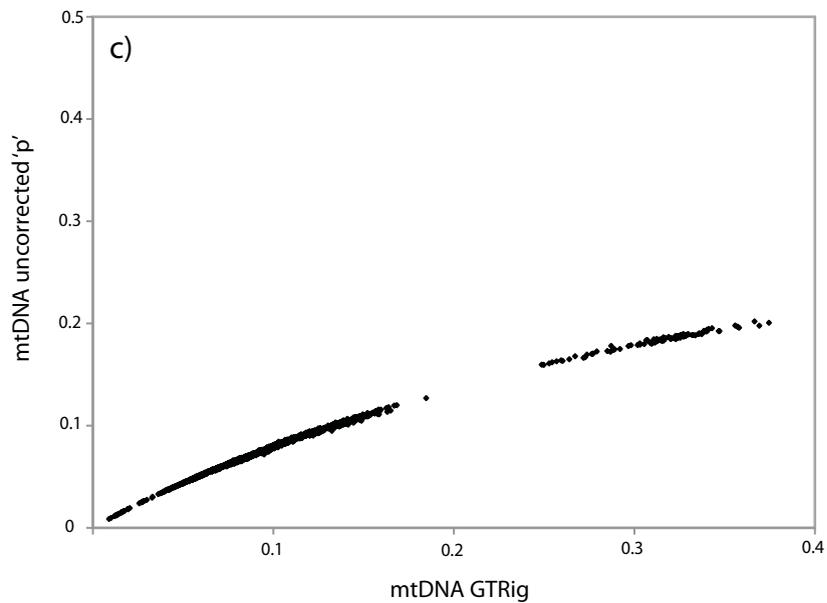
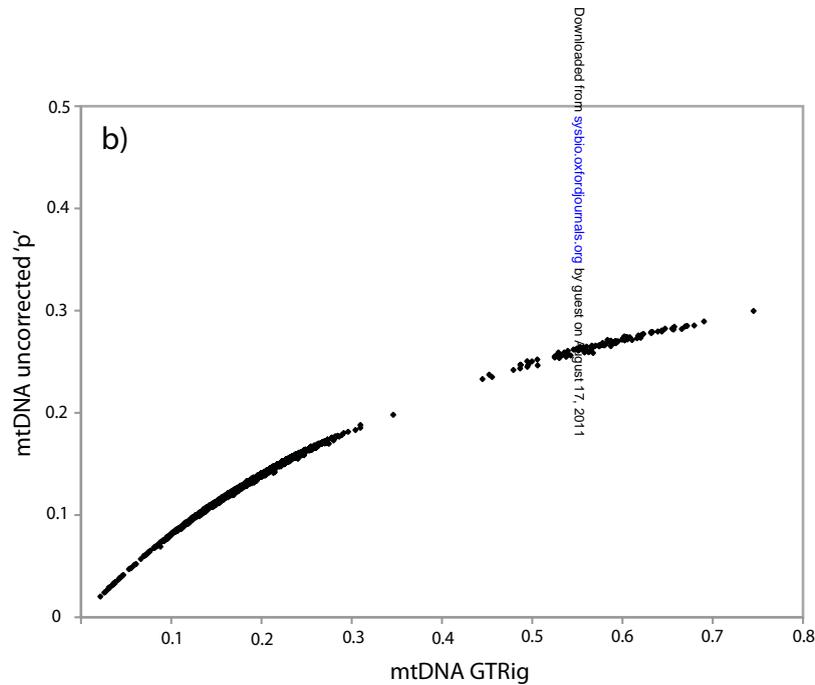
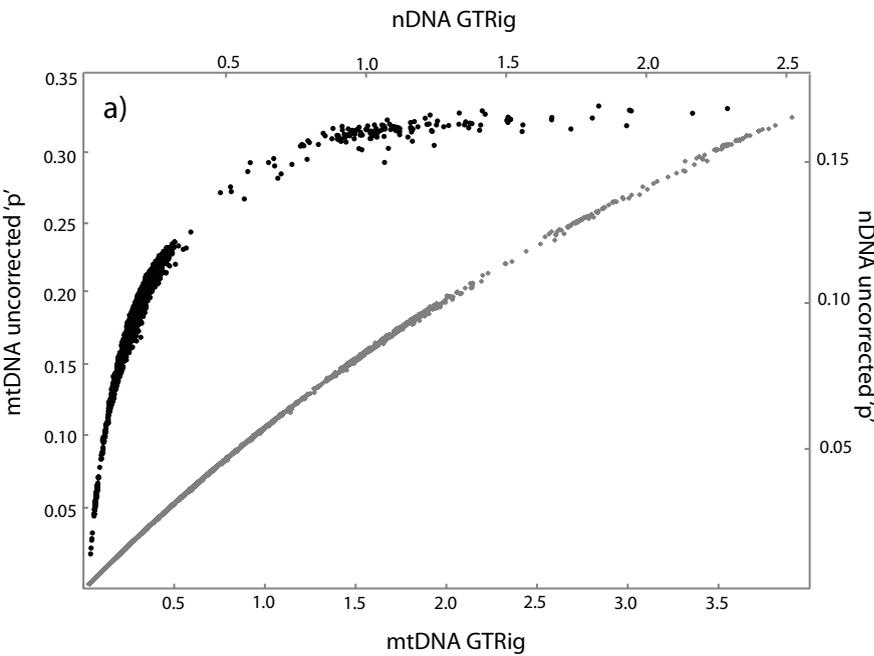
844

845 **Figure 5:** Effect of sequentially removing candidate fossil calibrated nodes on s , the average squared
 846 deviation of D_i values for the remaining fossil calibrations in each set. *5a:* Nuclear DNA s values for three
 847 calibration sets. Fossils were removed based on highest to lowest SS values calculated from all ten fossil
 848 calibrated nodes. Removal order (shown on the X -axis) of the first four most incongruent fossils was
 849 identical for each calibration set but then differed between sets. *5b:* MtDNA s values for three calibration
 850 sets when fossils were removed based on highest to lowest SS values calculated for all ten fossil calibrated
 851 nodes.

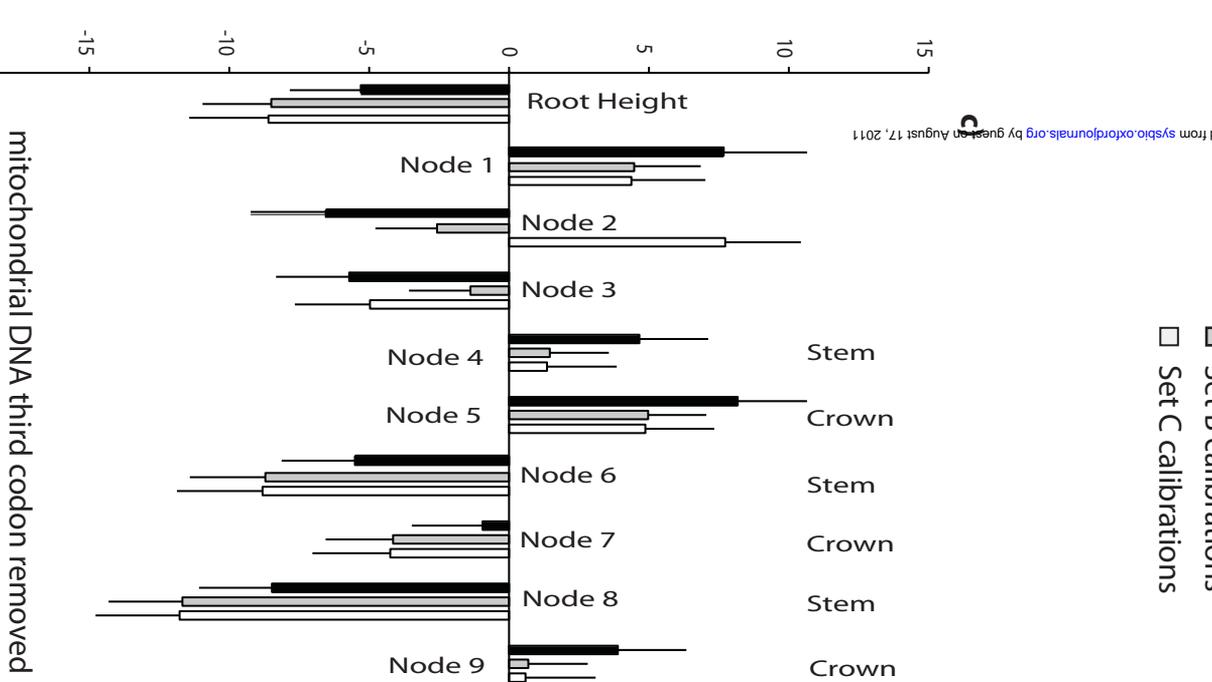
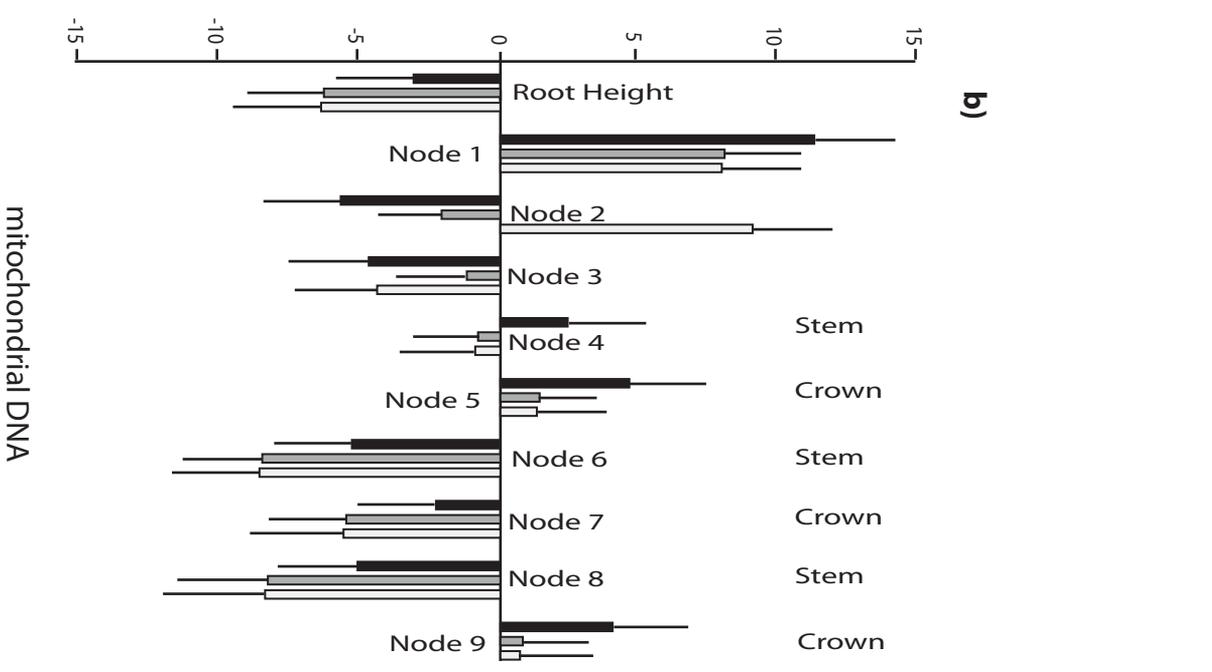
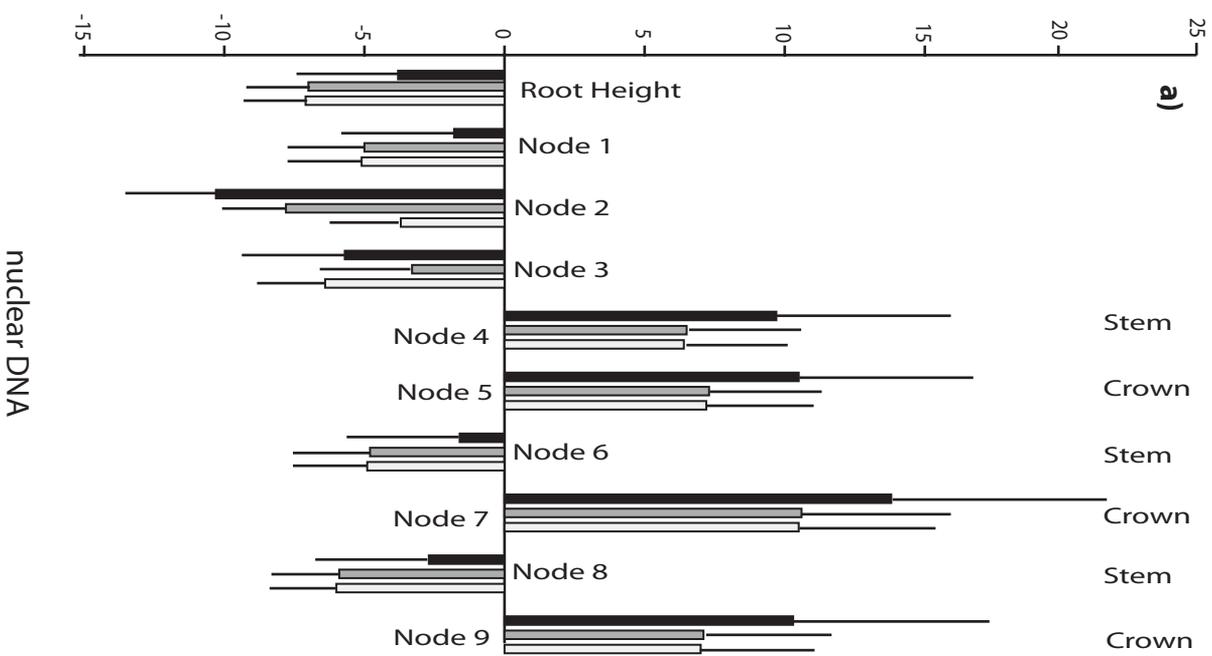
852

853 **Figure 6:** Bayesian multi-fossil calibration analyses showing fossil calibration priors and posterior
 854 distributions of molecular age estimates (mean and 95% HPD intervals) at six fossil calibrated nodes using
 855 four calibration sets (*A, B, C, D*). Each calibration set comprised four calibration priors that were identical
 856 among sets (tree root, nodes 1, 6 and 8) and two priors that differed among sets (nodes 2 and 3). Lognormal
 857 calibration priors are shown as wider shaded bars with the lognormal mean shown as a black square on the
 858 bar. Molecular age estimates for nuclear (black bars); mitochondrial (white bars) and mitochondrial DNA
 859 with third codon position removed (gray bars) are shown in pairs for each calibration set at each node. Bars
 860 indicate 95% HPDs with estimated mean ages indicated by black squares. At nodes 2 and 3 the respective
 861 calibration prior is shown immediately below the corresponding nuclear and mitochondrial age estimates.
 862 Calibration priors at the other four nodes are shown below all four sets of molecular age estimates. Prior
 863 and posterior distributions are shown on a diagrammatic chronogram depicting the backbone of the
 864 phylogeny, however this chronogram does not represent the results of any specific analysis.



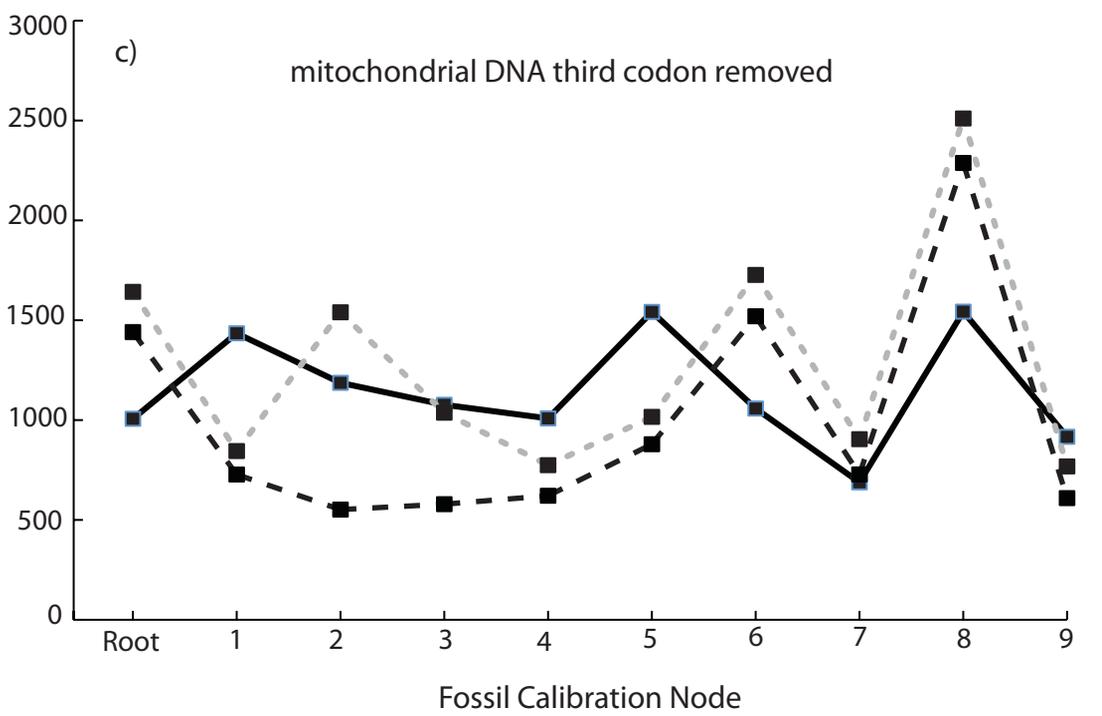
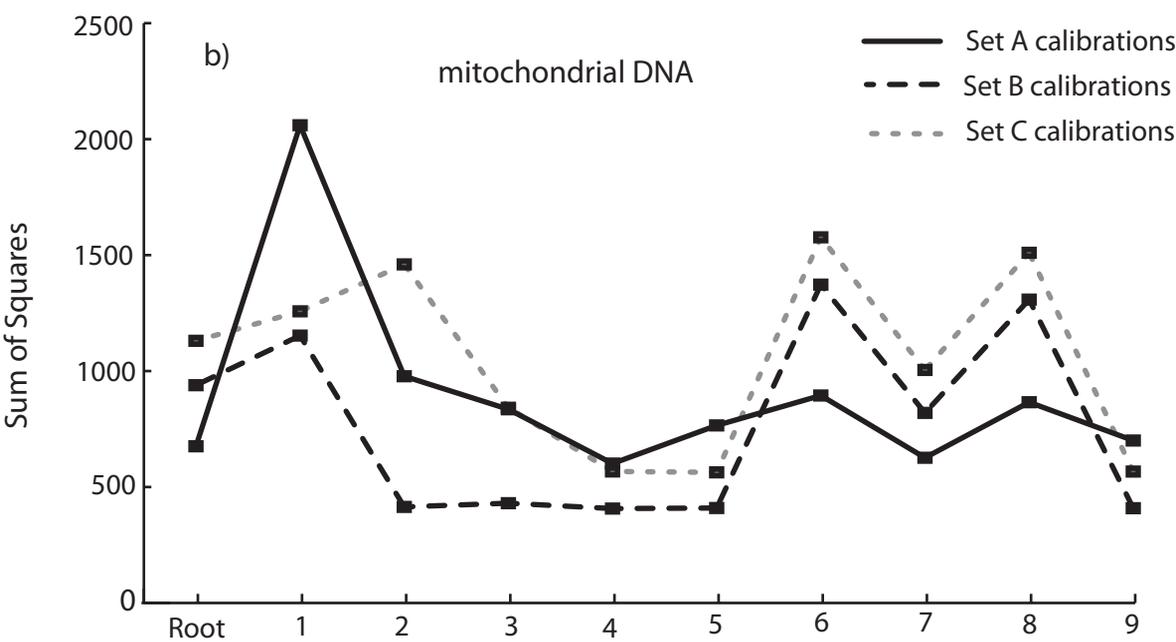
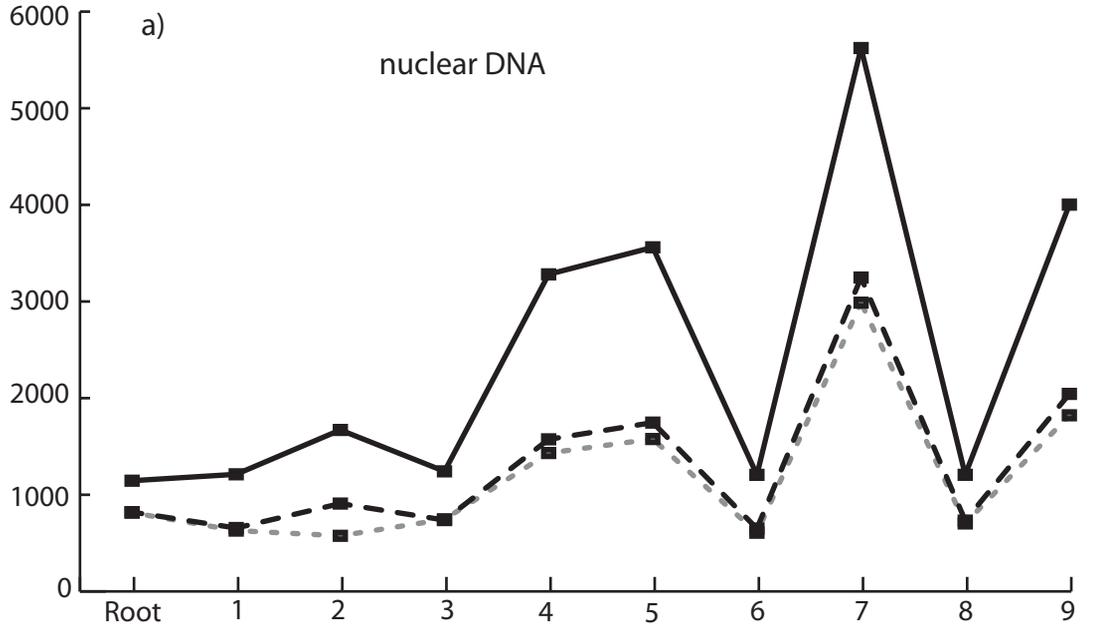


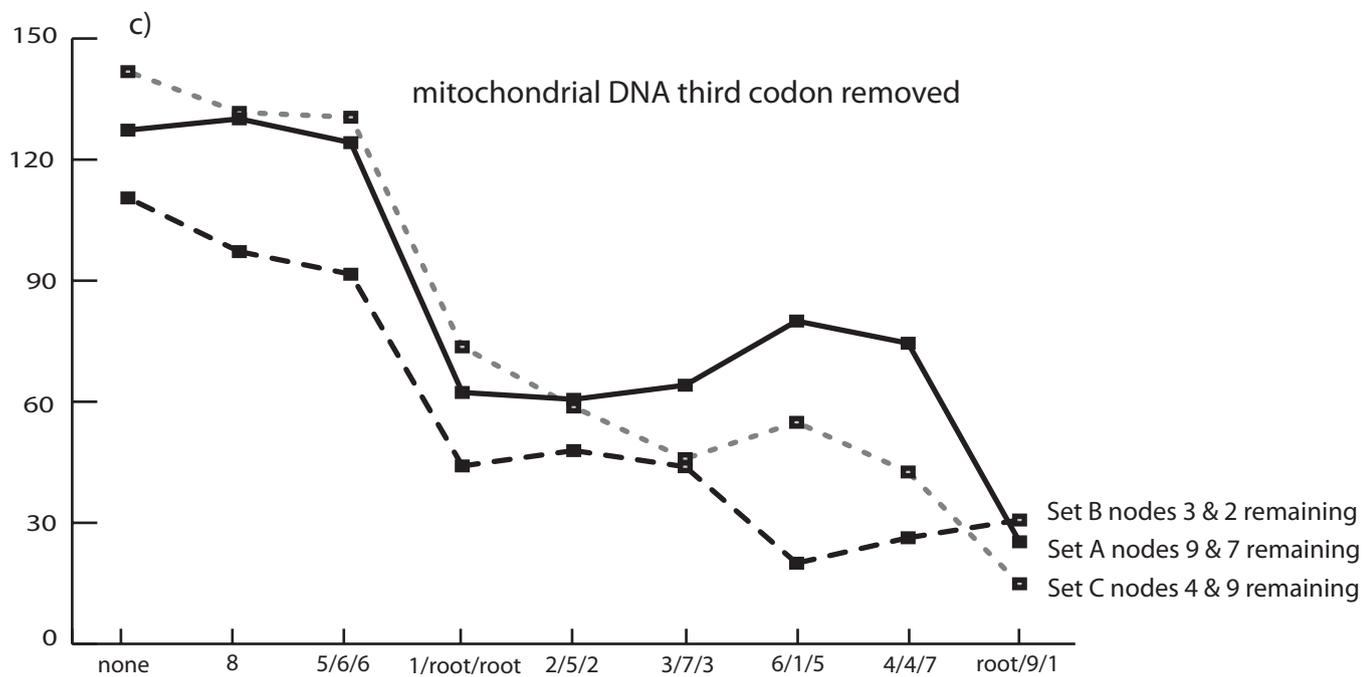
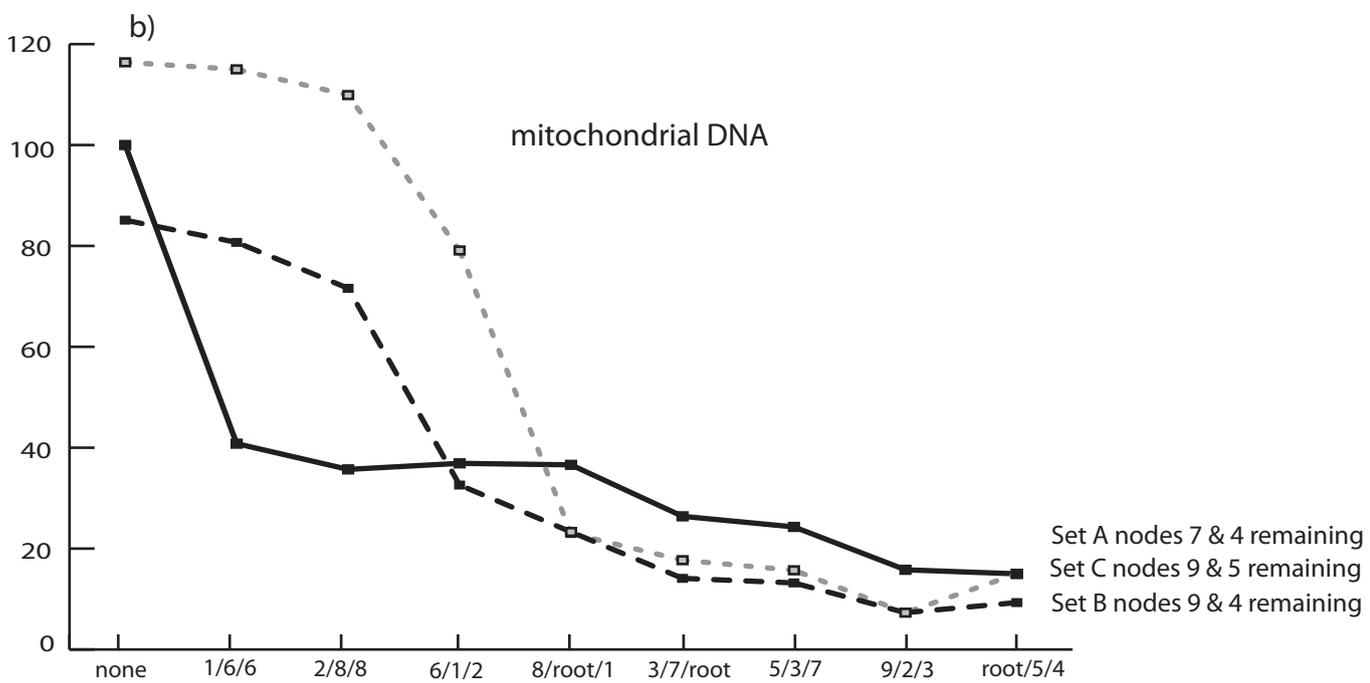
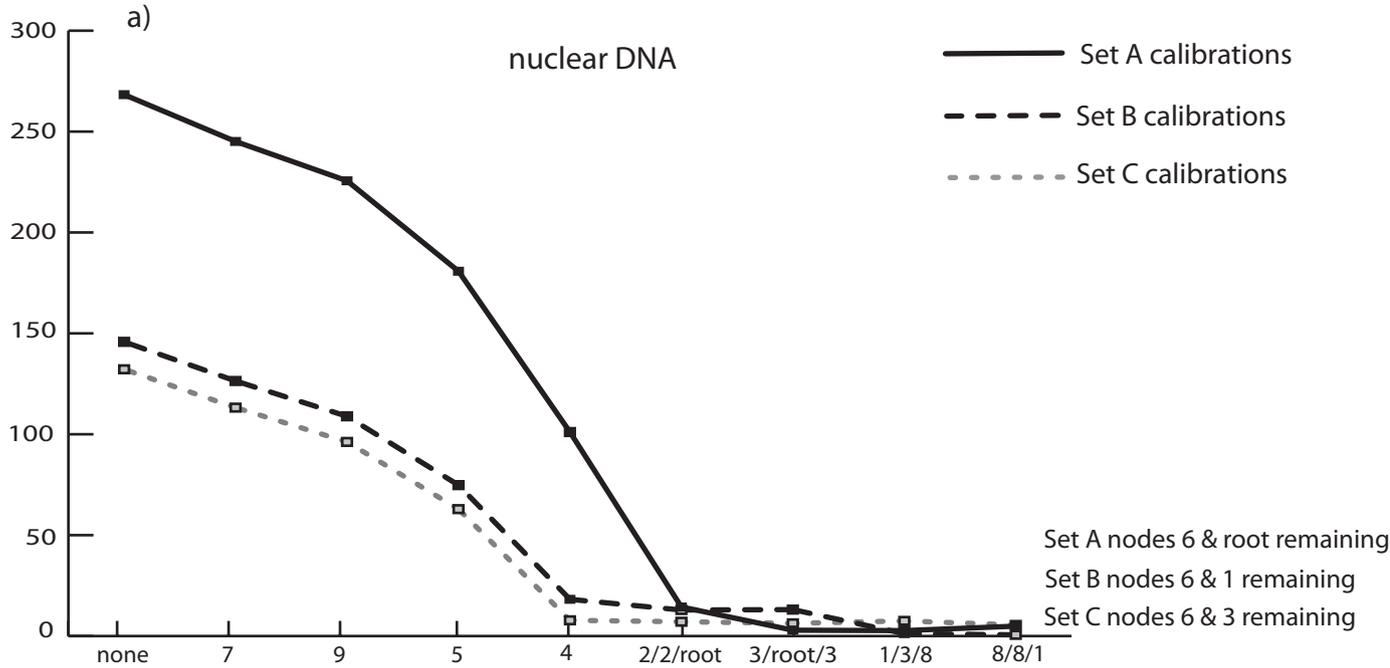
Mean difference between DNA ages & fossil dates (Myr)



Downloaded from <https://academic.oup.com/iob/advance-article-abstract/doi/10.1093/iob/obz017/5508881> by guest on August 17, 2019

Set A calibrations
 Set B calibrations
 Set C calibrations





Fossil Calibration Node Removed

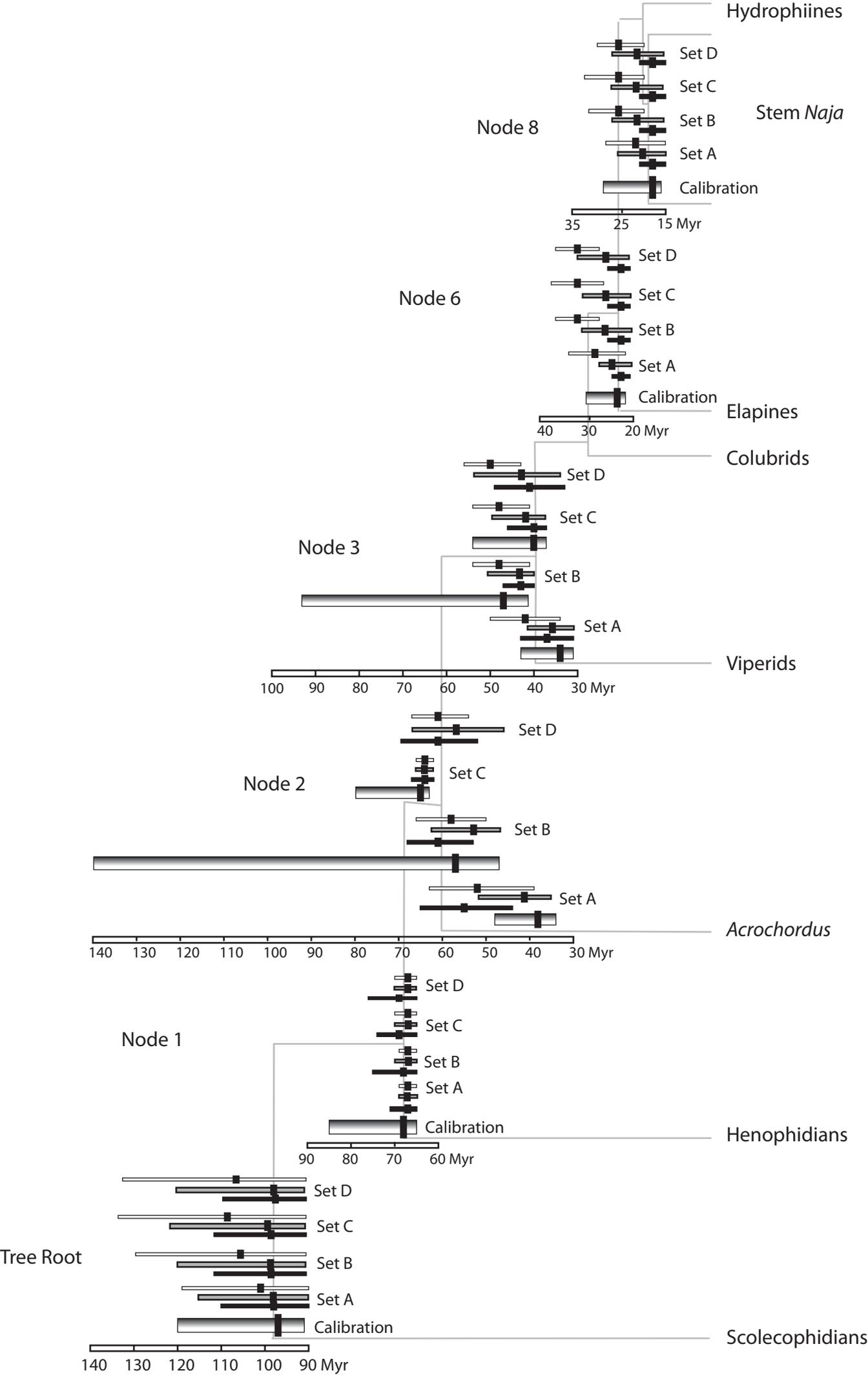


Table 1. Details of fossils tested using three approaches for evaluating candidate fossil calibrations. Constraints are given as absolute values (millions of years before present) and the corresponding lognormal mean, standard deviation and zero offset of the calibration prior used in BEAST analyses. Phylogenetic placement of nodes is shown on Figure 1.

Fossil Calibrations	Node	Calibration Priors			Reference
		Mean (95% HPD)	Ln Mean (stdev)	Zero offset	
Scolecophidians vs. alethinophidians	Root	97 (92-100)	2.00 (0.85)	90	The divergence between the Scolecophidia and the Alethinophidia was calibrated based on the earliest alethinophidian fossils: two <i>Coniophis</i> vertebrate from Utah from the upper Albanian / lowermost Cenomanian (97-102 Mya) (Gardner and Cifelli, 1999) and six <i>Coniophis</i> trunk vertebrate from the Cenomanian (94-100 Mya) in Sudan (Rage and Werner, 1999). Gardner and Cifelli (1999, pg. 95) note that the approximately contemporaneous occurrence of <i>Coniophis</i> fossils in geographically distant Sudan and Utah suggests that the Alethinophidia-Scolecophidia split occurred prior to the Cenomanian (99 Mya). This calibration also was used by (Kelly et al., 2009; Sanders and Lee, 2008).
Henophidians vs. caenophidians	1	68 (65-85)	1.00 (1.20)	65	The divergence between the Henophidia (booids) and Caenophidia (advanced snakes) has been dated using the fossils assigned to the Booidae. Noonan and Chippindale (2006a) dated the Henophidia-Caenophidia split at >75 Mya based on the earliest probable boid fossils from the latest Cretaceous (65-85 Mya) from South America. However the taxonomic affinities of these older vertebrae were not easy to assign (Albino, 2000; Rage, 2001). The first vertebrae that are undoubtedly booids occur in the mid-Palaeocene (58.5-56.5 Mya). These vertebrae are assigned to the extant genus <i>Corallus</i> (Boinae) and occur contemporaneously with fossil vertebrae from several other boine taxa (Rage, 2001) indicating that the Boinae were a separate phylogenetic entity by the mid-Palaeocene and that extant boine lineages originated early in the Tertiary or late Cretaceous (Rage et al. 2001, pg. 146). Based on these fossils we constrained the Henophidia-Caenophidia split as occurring 68 (65-85) Mya.
Acrochordids vs. colubroids	2 - Set A	38 (34-48)	1.40 (0.75)	34	The MRCA of the Caenophidia (Acrochordidae vs. Colubroidea) has been ascribed a range of dates based on different interpretations of the taxonomic affinities of certain fossils. These fossils include six vertebrae from the Cenomanian (93-96 Mya) in Sudan that were assigned to the Colubroidea (Rage and Werner, 1999); the oldest <i>Nigerophis</i> (Nigeropheidae) vertebra found in Paleocene marine deposits in Nigeria (56-65 Mya) (Rage, 1984, 1987); and the oldest undisputed colubroid fossil from the late-middle Eocene (37-39 Mya) (Head et al., 2005). We tested the effects of constraining this node with the three different divergence dates previously used based on these fossils: 38 (34-48) My (Kelly et al., 2009; Sanders and Lee, 2008); 57 (47-140) My (Wuster et al., 2008); and 65 (63-80) My (Noonan and Chippindale, 2006a, b) used > 65 My).
Acrochordids vs. colubroids	2 - Set B	57 (47-140)	2.50 (1.25)	45	
Acrochordids vs. colubroids	2 - Set C	65 (63-80)	1.10 (1.10)	62	
Viperids vs. colubrids + elapids	3 - Set A	34 (31-43)	1.40 (0.70)	30	The MRCA of the Colubroidea (viperids vs. colubrids and elapids) has been dated based on the oldest colubrid fossils from the Late Eocene (34-37 Mya) in Thailand (Rage et al., 1992), however, the oldest putative colubroid fossils from the Cenomanian (93-96 Mya) (Rage and Werner, 1999) have also been used to constrain the upper bound of this clade. Head et al., (2005) (pg.249) and Parmley and Holman, (2003) (pg. 6) argue that taxonomically and geographically divergent colubrid fossils found the late Eocene in Krabi Basin (Rage et al., 1992), Pondaung (Head et al., 2005) and North America (Parmley and Holman, 2003) indicate that colubroids had started diverging pre-Late Eocene, possibly even in the early Paleogene
Viperids vs. colubrids + elapids	3 - Set B	47 (40-95)	2.00 (1.20)	40	

Viperids vs. colubrids + elapids	3 - Set C	40 (37-60)	1.10 (1.25)	37	(43-60 Mya). We tested two divergence dates previously used: 34 (31-43) My (Kelly et al., 2009; Sanders and Lee, 2008; Wiens et al., 2006); and 47 (40-95) My (Wuster et al., 2008). We also tested a constraint of 40 (37-60) My based on the geographically and taxonomically divergent colubrid fossils from the Late Eocene assigned to <i>Coluber cadureci</i> and <i>Natrix mlynarskii</i> , extinct species that belong the extant sub-
Natricines vs. colubrids (Stem)	4	36 (35-45)	0.50 (1.10)	35	families Colubrinae and Natricinae respectively, have been described from the early Oligocene (30-34 Mya) in Europe (Rage, 1988). A third colubrid, <i>Texasophis galbreathii</i> , has been described from the early Orellan to Whitneyan ages of the Oligocene (30-31 Mya) in North America (Holman, 1984) (pg. 225). Based on these fossils the crown natricine-colubrine divergence has been constrained at 35-45 My (Guicking et al., 2006; Alfaro et al., 2008). However, fossils with colubrine and natricine morphology appear almost immediately after the first appearance of indeterminate colubrids, suggesting that these primitive fossils may be more appropriate for dating the stem natricine-colubrine clade (in our case the divergence between the xenodontines, natricines, colubrids). We tested the effect of constraining the stem (node 4) and crown (node 5) colubrine-natricine clades at 37 (35-45) My.
Natricines vs. colubrids (Crown)	5				
Elapines vs. hydrophiines	6	23 (21-30)	1.00 (0.80)	20	A fossil vertebra from the late Oligocene/early Miocene (20-23 Mya) has been assigned to <i>Laticauda</i> and, based on its similarity to <i>L. colubrina</i> but differences from <i>L. laticaudata</i> and other elapids, Scanlon et al., (2003) (pg. 579) suggested that this fossil is nested within (not basal to) the genus <i>Laticauda</i> . Based on this taxonomic assignment, Wuster et al. (2007) used a minimum age of 24 My to calibrate the divergence between <i>Laticauda</i> and all other hydrophiines (crown hydrophiines). However, the taxonomic affinity and/or stratigraphic age of this fossil have recently been questioned (Sanders & Lee, 2008, pg. 1186). This vertebra is one of the oldest elapid fossil known and might, therefore, be basal to (rather than nested within) the extant elapid group. Apart from this fossil, the earliest appearances of modern elapids in the first fossil record are proteroglyphous fangs from Germany dated at 20-23 Mya (Kuch et al., 2006). We tested the effects of constraining the crown hydrophiines and crown elapids with dates of 23 (21-30) My.
Crown hydrophiines	7				
African vs. Asian <i>Naja</i> (Stem)	8	19 (17-30)	1.00 (1.00)	16	Fossils of three extinct European <i>Naja</i> species with apomorphies that distinguish Asian and African <i>Naja</i> occur 16 Mya (Szyndlar and Rage, 1990). These fossils have been used to date the divergence between the crown African and Asian <i>Naja</i> (Kelly et al., 2009; Wuster et al., 2007, 2008). However, these extinct fossil species display primitive conditions that are very rare among living cobras (Szyndlar and Rage, 1990) suggesting that they should be used to calibrate the stem rather than the crown <i>Naja</i> clade. We assigned the divergence between <i>Naja</i> and the closely related <i>Bungarus</i> as the stem clade and explored the effects of constraining the crown and stem <i>Naja</i> with dates of 19 (17-30) My.
African vs. Asian <i>Naja</i> (Crown)	9				

Table 2. Empirical Scaling Factors (ESF_i) for candidate fossil calibrations for nuclear and mitochondrial datasets calculated using proportional branch lengths obtained from uncalibrated ultrametric trees produced using different methods. The BEAST nDNA chronogram was obtained by fixing the root to an arbitrary value of 100. The remaining ultrametric trees were produced in r8s using the optimal maximum likelihood (ML) and Bayesian (MrBayes) phylogenies. Uncalibrated ultrametric trees were obtained by fixing the root to an arbitrary value of 100 and using penalised likelihood with the logarithmic (log) or additive (add) penalty function and the optimal smoothing parameter obtained from cross-validation (shown in the column heading). Nodes and corresponding ESF values highlighted in bold for each ultrametric tree indicate the fossil with the highest empirical coverage after removing fossils identified as outliers (i.e. not conforming to a uniform distribution). See text for more details.

Nuclear DNA						Mitochondrial DNA				mtDNA noThird			
MrBayes - log 3200		ML - add 3200		ML - log 320		BEAST		MrBayes - log 1		ML - log 10		ML - log 10	
Node #	ESF	Node #	ESF	Node #	ESF	Node #	ESF	Node #	ESF	Node #	ESF	Node #	ESF
2 - Set A	76	2 - Set A	79	2 - Set A	86	2 - Set A	57	Root	97	Root	97	Root	97
Root	97	Root	97	Root	97	3 - Set A	72	2 - Set A	148	2 - Set A	165	2 - Set A	128
3 - Set A	113	3 - Set A	115	2 - Set B	130	3 - Set C	84	3 - Set A	178	3 - Set A	193	3 - Set A	137
2 - Set B	114	2 - Set B	118	1	138	1	85	3 - Set C	209	3 - Set C	227	8	140
8	122	8	120	3 - Set A	140	2 - Set B	85	1	213	8	232	6	148
1	123	1	128	2 - Set C	148	6	85	8	221	1	241	3 - Set C	161
2 - Set C	130	2 - Set C	135	3 - Set C	165	8	88	2 - Set B	223	6	245	4	185
3 - Set C	132	3 - Set C	136	8	168	2 - Set C	97	6	235	2 - Set B	247	7	187
6	137	6	152	3 - Set B	193	Root	97	3 - Set B	246	4	265	3 - Set B	190
3 - Set B	156	3 - Set B	159	6	224	3 - Set B	99	4	247	3 - Set B	267	2 - Set B	192
4	196	4	194	4	250	7	130	2 - Set C	254	2 - Set C	281	1	197
5	196	5	194	5	250	4	134	5	271	5	286	9	203
7	235	7	242	7	361	9	168	7	299	7	324	5	205
9	250	9	370	9	494	5	171	9	373	9	404	2 - Set C	218

1Lukoschek et al., *Supplementary Material A*.

2*Evaluating Fossil Calibrations*

3*Cross-validations and empirical coverage*

4Nuclear DNA identified the crown placements of the *Naja* and *Laticauda* fossils as outliers using
5both the cross-validation analyses and empirical scaling factors; however, this was not the case
6for the mitochondrial data. We accounted for these differences in terms of the effects of
7mitochondrial saturation, however, it is also important to evaluate these results in terms of the
8fossils themselves.

9

10The *Naja* fossils comprise three extinct species with characters that distinguish Asian and
11African *Naja* (Szyndlar and Rage, 1990). However, these fossils also have primitive characters
12very rare among living cobras (Szyndlar and Rage, 1990, pg 398) suggesting that they belong to
13the stem *Naja*. When used to constrain the divergence between extant African and Asian *Naja*
14species (crown *Naja* - node 9) (Kelly et al., 2009; Wuster et al., 2007; Wuster et al., 2008), these
15fossils overestimated dates for other fossil-calibrated nodes (Fig. 2a) and were identified as
16outliers by empirical scaling factors (Table 2). However, our alternative placement for dating the
17*Naja-Bungarus* divergence (node 8), based on previous evidence of the close relationships
18between *Naja* and *Bungarus* (Slowinski and Keogh, 2000; Wuster et al., 2007), consistently
19produced much younger divergence dates (Fig. 2) and had intermediate empirical coverage
20(Table 2). *Naja* is paraphyletic with *Boulengerina* and *Paranaja* and its relationships with other
21cobra genera are poorly resolved (Slowinski and Keogh, 2000; Wuster et al., 2007); thus, its
22sister group is difficult to identify as is the most appropriate placement of this fossil on the tree.
23However, the *Naja* fossils include well-preserved skull elements with well-defined
24morphological characters (Szyndlar and Rage, 1990); thus, the best nodal placement might be
25identified from cladistic analysis of extinct and extant taxa (Doyle and Donoghue, 1993). The
26relative completeness of these fossils also means they could be used in a Bayesian approach that
27incorporates morphological data from fossils and extant species (Lee et al., 2009) to evaluate the

28effects of alternative nodal placements on estimated divergence times.

29

30An elapid fossil from the Australian late Oligocene/early Miocene (20-23 MA) also consistently
31overestimated dates at other fossil calibrated nodes when used to calibrate the crown
32hydrophiines (Fig. 2a - node 7) and was identified as an outlier by empirical scaling factors
33(Table 2). This juvenile vertebra was described as being nested within *Laticauda* rather than
34within or basal to any other elapid clade (Scanlon et al., 2003); however, there have been calls for
35taxonomic and/or stratigraphic revisions of this fossil (J. Scanlon pers. comm.), partly in
36response to younger molecular dates obtained for the divergence between *Laticauda* and the
37remaining hydrophiines (Sanders and Lee, 2008). Our analyses clearly are unable to resolve
38either the taxonomic affinities or correct stratigraphy of this fossil. Nonetheless, the alternative
39placement of this fossil to constrain the crown elapids (node 6) tended to produce much younger
40estimates of other fossil dates (Fig. 2a), though it had amongst the highest empirical scaling
41factors (Table 2). Proteroglyphous fangs very similar to those of modern elapids first appear in
42the fossil record in Germany 20-23 MA (Kuch et al., 2006) suggesting that earlier constraints
43probably might be more realistic for the crown elapids.

44

45The nuclear DNA cross-validations indicated that constraining the stem and crown natricine-
46colubrine clades (nodes 4 & 5 respectively) at 36 (35-45) My (Alfaro et al., 2008; Guicking et al.,
472006) overestimated dates at other fossil calibrated nodes (Fig. 3a). This result is not surprising
48given that the natricine-colubrine divergence is much shallower than the crown Colubroidea
49(node 3) in the nuclear gene tree (Figs. 1a), yet their calibrations overlapped to greater or lesser
50extents (Table 1). Indeed, the colubrine-natricine constraint was almost identical to the *set A*
51constraint for the MRCA of all advanced snakes (Table 1 - node 2). This discrepancy is not
52confined to our study: Alfaro et al., (2008) used 35-55 MA to constrain the tree-root (crown
53colubroids) and 35-45 MA for the deeply nested natricine-colubrine divergence (see their Fig. 2).
54These overlapping constraints reflect the fact that the oldest known colubrine fossil, a vertebra

55 assigned to *Nebraskophis* from the Late Eocene in North America (Parmley and Holman, 2003),
56 occurs simultaneously with the oldest undoubted colubrid fossil from the late Eocene in Thailand
57 (34-37 MA) (Rage et al., 1992), while earliest natricine fossil (*Natrix mlynarskii*) appears in
58 Europe soon after in the early Oligocene (32-34 MA), where it co-occurs with colubrine
59 (*Coluber cadurci*) fossils (Rage, 1988). The nodal age for the crown Colubroidea has been
60 inferred from the oldest undoubted colubrid fossil (see below), while the colubrine-natricine
61 divergence has been inferred from the oldest natricine fossil. However molecular phylogenetic
62 appraisals invariably infer nested positions for the colubrines and natricines among the viperids,
63 elapids, atractaspids, and other colubrid subfamilies that comprise the Colubroidea (Kelly et al.,
64 2009; Lawson et al., 2005; Vidal et al., 2007; Yan et al., 2008; Zaher et al., 2009).

65

66 How can these discrepancies be resolved? Firstly, Rage (1988, pg. 467) questioned the
67 taxonomic affinities of the natricine fossil in its description. Apart from this fossil, the next
68 appearance of natricine morphology in the fossil record does not occur until the early Miocene
69 (20-23 MA), when several natricine species appear (Rage and Auge, 1993; Ivanov, 2001). The
70 earliest 'natricine' fossil may, therefore, be so deeply buried in the stem lineage as to be
71 irrelevant for dating the divergence between modern colubrines and natricines. Secondly,
72 molecular phylogenies do not resolve the natricines and colubrines as sister taxa. Instead, basal
73 divergences among extant colubrid clades comprise one or more poorly resolved polytomies
74 (Lawson et al., 2005; Vidal et al., 2007; Kelly et al., 2009) and a sister-group relationship among
75 colubrines and natricines is only recovered if other clades in the polytomy are not sampled (as in
76 our study and also in Alfaro et al., 2008). Thus, even if the Oligocene fossil is a natricine, its
77 appropriate placement is deeper in the tree. Finally, the first appearance of an undoubted
78 colubrid probably does not represent the earliest divergences among the Colubroidea. Molecular
79 phylogenetic hypotheses indicate that other clades (e.g. viperids, homalopsids, pareatids)
80 diverged earlier in the colubroid radiation (Lawson et al., 2005; Vidal et al., 2007; Alfaro et al.,
81 2008; Wuster et al., 2008); however, the earliest known fossils from these clades date to the early

82Miocene (Rage and Auge, 1993; Szyndlar and Rage, 1999; Ivanov, 2001).

83

84*Multi-calibration Bayesian evaluation of node 2 and 3 calibration sets*

85The alternative calibration for nodes 2 and 3 were evaluated by comparing the prior and posterior
86distributions from multi-calibration Bayesian analyses (Fig. 3). The lognormal priors we used
87model the probability distribution of the actual emergence dates of a clade using the fossil date to
88inform the hard minimum bound for the youngest possible age, specifying a mean age somewhat
89older than the fossil, and have a soft maximum bound that allows for the clade to be considerably
90older than the fossil record (Ho and Phillips, 2009; Yang and Rannala, 2006). It is important to
91note that the maximum bound of a lognormal prior accounts for a relatively small proportion of
92the total prior probability and thus has a much smaller effect on the posterior distribution than the
93log-normal mean or mode, which accounts for a much larger proportion of the prior probability
94and thus exerts a far stronger constraint (see Fig. 2f in Ho and Phillips 2009). Moreover, the hard
95minimum bound gives zero probability to the nodal age actually being younger than the oldest
96fossil known (Ho and Phillips, 2009). As such, posterior distributions that are younger than their
97lognormal calibration priors strongly suggest that their respective calibrations were too old.

98

99Comparing the *set B* posterior distributions with their respective priors suggests that the *set B*
100calibrations are too old (Fig. 3). In particular, the estimated maximum 95% HPDs are tens of
101millions of years younger than their respective calibration priors, with the mean and minimum
102bound for the crown colubroids (node 3) also younger than its calibration prior (Fig. 3). By
103contrast, the *set A* posterior distribution for the crown caenophidians (node 2) was much older
104than its prior (in fact they barely overlapped) indicating that the *set A* calibration was too young
105(Fig. 3). The young *set A* constraint of 38 (34-48) My for crown caenophidians (node 2) (Kelly
106et al., 2009; Sanders and Lee, 2008) was based on the first appearance of undisputed colubroids
107(37-39 My) in the fossil record (Head et al., 2005). However, these same fossils also were used
108to constrain the minimum bound of the crown colubroids (*set B* - node 3) at 40 My (Wuster et al.,

1092008), with an maximum bound of 95 My based on possible colubroid fossils from the
110Cenomanian (93-96 MA) (Rage and Werner, 1999). They were then further used to extrapolate
111even older dates of 57 (47-140) MA for the origins of the Caenophidia (*set B* – node 2) (Wuster
112et al., 2008). Ignoring the contentious taxonomic affinities of the oldest Cenomanian colubroid
113fossils (Head et al., 2005; Sanders and Lee, 2008), the first primitive colubroids to appear in the
114fossil record almost certainly belong to the stem colubroids; thus, are inappropriate for dating the
115crown colubroid radiation (node 3) (Doyle and Donoghue, 1993; Magallon and Sanderson,
1162001).

117

118The *set C* estimated mean and minimum HPD for caenophidian origins (node 2) also were
119younger than its calibration prior of 65 (63-80) MA (Noonan and Chippindale, 2006a; Noonan
120and Chippindale, 2006b). This constraint was based on the first *Nigerophis* fossil dated at 56-65
121My (Rage, 1984). However, the relevance of this fossil for dating caenophidian origins relies on
122the *Nigeropheidae* belonging to the *Acrochordoidea* (McDowell, 1987; Rage, 1987) but
123morphological similarities of this fossil to several disparate groups renders its taxonomic
124affinities uncertain (Rage, 1984, pg. 71). The maximum bound of this calibration ignores an
125indeterminate colubroid fossil dated at 49-56 My (Rage et al., 2003) that suggests the
126Caenophidia started diverging in the early to mid-Eocene. (Head et al., 2005) suggested that this
127‘colubroid’ fossil might be an acrochordid: nonetheless, even if reassigned the fossil remains
128relevant for dating caenophidian origins (acrochordoid-colubroid divergence - node 2).

129

130

131Alfaro, M. E., D. R. Karns, H. H. Voris, C. D. Brock, and B. L. Stuart. 2008. Phylogeny,
132 evolutionary history, and biogeography of Oriental-Australian rear-fanged water snakes
133 (Colubroidea: Homalopsidae) inferred from mitochondrial and nuclear DNA sequences.
134 Mol Phylogenet Evol 46:576-593.

135Doyle, J. A., and M. J. Donoghue. 1993. Phylogenies and angiosperm diversification.

- 136 Paleobiology 19:141-167.
- 137Guicking, D., R. Lawson, U. Joger, and M. Wink. 2006. Evolution and phylogeny of the genus
138 *Natrix* (Serpentes: Colubridae). Biol J Linn Soc 87:127-143.
- 139Head, J., P. A. Holroyd, J. H. Hutchison, and R. L. Ciochon. 2005. First report of snakes
140 (Serpentes) from the late middle Eocene Pondaung formation, Myanmar. J Vertebr
141 Paleontol 25:246-250.
- 142Ho, S. Y. W., and M. J. Phillips. 2009. Accounting for calibration uncertainty in phylogenetic
143 estimation of evolutionary divergence times. Syst Biol 58:367-380.
- 144Kelly, C. M. R., N. P. Barker, M. H. Villet, and D. G. Broadley. 2009. Phylogeny, biogeography
145 and classification of the superfamily Elapoidea: a rapid radiation in the late Eocene.
146 Cladistics 25:38-63.
- 147Kuch, U., J. Muller, C. Modden, and D. Mebs. 2006. Snake fangs from the lower Miocene of
148 Germany; evolutionary stability of perfect weapons. Naturewissenschaften 93:84-87.
- 149Lawson, R., J. B. Slowinski, B. I. Crother, and F. R. Burbrink. 2005. Phylogeny of the
150 Colubroidea (Serpentes): New evidence from mitochondrial and nuclear genes. Mol
151 Phylogenet Evol 37:581-601.
- 152Lee, M. S. Y., P. G. Oliver, and M. N. Hutchison. 2009. Phylogenetic uncertainty and molecular
153 clock calibrations: a case study of legless lizards (Pygopodidae, Gekkota). Mol
154 Phylogenet Evol 50:661-666.
- 155Magallon, S., and M. J. Sanderson. 2001. Absolute diversification rates in angiosperm clades.
156 Evolution 55:1762-1780.
- 157McDowell, J. R. 1987. Systematics. Pages 3-50 in *Snakes - ecology and evolutionary biology* (R.
158 A. Seigel, J. T. Collins, and S. S. Novak, eds.). Macmillan, New York.
- 159Noonan, B. P., and P. T. Chippindale. 2006a. Dispersal and vicariance: the complex evolutionary
160 history of boid snakes. Mol Phylogenet Evol 40:347-358.
- 161Noonan, B. P., and P. T. Chippindale. 2006b. Vicariant origin of Malagasy reptiles supports late
162 Cretaceous Antarctic land bridge. The American Naturalist 168:730-741.

- 163Parnley, D., and J. A. Holman. 2003. *Nebraskophis* HOLMAN from the Late Eocene of Georgia
164 (USA), the oldest known North American colubrid snake. *Acta Zool Cracov* 46:1-8.
- 165Rage, J. C. 1984. Serpentes. Pages 1-79 in *Encyclopedia of Paleoherpétology* Gustav Fischer
166 Verlag, Stuttgart.
- 167Rage, J. C. 1987. Fossil history. Pages 51-76 in *Snakes - ecology and evolutionary biology* (R. A.
168 Seigel, J. T. Collins, and S. S. Novak, eds.). Macmillan, New York.
- 169Rage, J. C., S. Bajpai, J. G. M. Thewissen, and B. N. Tiwari. 2003. Early Eocene snakes from
170 Kutch, Western India, with a review of the Palaeophiidae. *Geodiversitas* 25:
171695-716.
- 172Rage, J. C., E. Buffetaut, H. Buffetaut-Tong, Y. Chaimanee, S. Ducrocq, J. J. Jaeger, and V.
173 Suteethorn. 1992. A colubrid in the late Eocene of Thailand: The oldest known
174 Colubridae (Reptilia, Serpentes). *Comptes Rendus de l'Académie des sciences* 314:1085-
175 1089.
- 176Rage, J. C., and C. Werner. 1999. Mid-Cretaceous (Cenomanian) snakes from Wadi Abu
177 Hashim, Sudan: The earliest snake assemblage. *Palaeontol Afr* 35:85-110.
- 178Sanders, K. L., and M. Y. L. Lee. 2008. Molecular evidence for a rapid late-Miocene radiation of
179 Australasian venomous snakes (Elapidae: Colubroidea). *Mol Phylogenet Evol* 46:1180-
180 1188.
- 181Scanlon, J. D., M. S. Y. Lee, and M. Archer. 2003. Mid-Tertiary elapid snakes (Squamata,
182 Colubroidea) from Riversleigh, northern Australia: early steps in a continent-wide
183 adaptive radiation. *Geobios* 36:573-601.
- 184Slowinski, J. B., and J. S. Keogh. 2000. Phylogenetic relationships of elapid snakes based on
185 cytochrome *b* mtDNA sequences. *Mol Phylogenet Evol* 15:157-164.
- 186Szyndlar, Z., and J. C. Rage. 1990. West Palearctic cobras of the genus *Naja* (Serpentes:
187 Elapidae): interrelationships among extinct and extant species. *Amphib-Reptilia* 11:385-
188 400.
- 189Vidal, N., A. S. Delmas, P. David, C. Cruaud, A. Couloux, and S. B. Hedges. 2007. The

190 phylogeny and classification of caenophidian snakes inferred from seven nuclear protein-
191 coding genes. *C R Biol* 330:182-187.

192Wuster, W., S. Crookes, I. Ineich, Y. Mane, C. E. Pook, J.-F. Trape, and D. G. Broadley. 2007.
193 The phylogeny of cobras inferred from mitochondrial DNA sequences: evolution of
194 venom spitting and the phylogeography of the African spitting cobras (Serpentes:
195 Elapidae: *Naja nigricollis* complex). *Mol Phylogenet Evol* 45:437-453.

196Wuster, W., L. Peppin, C. E. Pook, and D. E. Walker. 2008. A nesting of vipers: phylogeny and
197 historical biogeography of the Viperidae (Squamata: Serpentes). *Mol Phylogenet Evol*
198 49:445-459.

199Yan, J., H. Li, and K. Zhou. 2008. Evolution of the mitochondrial genome in snakes: gene
200 rearrangements and phylogenetic relationships. *BMC Genet* 9:569.

201Yang, Z., and B. Rannala. 2006. Bayesian estimation of species divergence times under a
202 molecular clock using multiple fossil calibrations with soft bounds. *Mol Biol Evol*
203 23:212-226.

204Zaher, H., F. G. Grazziotin, J. E. Cadle, R. W. Murphy, J. C. de Moura-Leite, and S. L. Bonatto.
205 2009. Molecular phylogeny of advanced snakes (Serpentes, Caenophidai) with an
206 emphasis on South American Xenodontines: a revised classification and descriptions of
207 new taxa. *Papeis Avulsos de Zoologia* 49:115-153.

208

209