

RH: BAYESIAN ESTIMATION OF CLADE AGES

# Bayesian Phylogenetic Estimation of Clade Ages Supports Trans-Atlantic Dispersal of Cichlid Fishes

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*Abstract.*— Divergence-time estimation based on molecular phylogenies and the fossil  
record has provided insights into fundamental questions of evolutionary biology. In  
Bayesian node dating, phylogenies are commonly time calibrated through the specification

21 of calibration densities on nodes representing clades with known fossil occurrences.  
22 Unfortunately, the optimal shape of these calibration densities is usually unknown and they  
23 are therefore often chosen arbitrarily, which directly impacts the reliability of the resulting  
24 age estimates. As possible solutions to this problem, two non-exclusive alternative  
25 approaches have recently been developed, the “fossilized birth-death” model and  
26 “total-evidence dating”. While these approaches have been shown to perform well under  
27 certain conditions, they require including all (or a random subset) of the fossils of each  
28 clade in the analysis, rather than just relying on the oldest fossils of clades. In addition,  
29 both approaches assume that fossil records of different clades in the phylogeny are all the  
30 product of the same underlying fossil sampling rate, even though this rate has been shown  
31 to differ strongly between higher-level taxa. We here develop a flexible new approach to  
32 Bayesian age estimation that combines advantages of node dating and the fossilized  
33 birth-death model. In our new approach, calibration densities are defined on the basis of  
34 first fossil occurrences and sampling rate estimates that can be specified separately for all  
35 clades. We verify our approach with a large number of simulated datasets, and compare its  
36 performance to that of the fossilized birth-death model. We find that our approach  
37 produces reliable age estimates that are robust to model violation, on par with the  
38 fossilized birth-death model. By applying our approach to a large dataset including  
39 sequence data from over 1000 species of teleost fishes as well as 147 carefully selected fossil  
40 constraints, we recover a timeline of teleost diversification that is incompatible with  
41 previously assumed vicariant divergences of freshwater fishes. Our results instead provide  
42 strong evidence for trans-oceanic dispersal of cichlids and other groups of teleost fishes.  
43 (Keywords: Bayesian inference; phylogeny; calibration density; relaxed molecular clock;  
44 fossil record; Cichlidae; marine dispersal)

45 In phylogenetic analyses, molecular sequence data are commonly used to infer not  
46 only the relationships between species, but also the divergence times between them. The  
47 estimation of divergence times in a phylogenetic context is usually based on an assumed  
48 correlation between the age of species separation and the number of observed genetic  
49 differences, i.e. a “molecular clock” (Zuckerlandl and Pauling 1962). Evidence for the  
50 existence of molecular clocks initially derived from relative rate tests (Sarich and Wilson  
51 1967) and has since been corroborated by a large body of literature (e.g. Wilson et al.  
52 1977; Bromham and Penny 2003). However, it has been shown that the rate of the  
53 molecular clock often differs between lineages (Drummond et al. 2006) and that it can  
54 depend on factors including body size, metabolic rate, and generation time (Martin and  
55 Palumbi 1993; Nabholz et al. 2008).

56 To allow the estimation of absolute divergence dates from sequence data, a  
57 calibration of the rate of the molecular clock is required. This calibration can be obtained  
58 from serially sampled DNA sequences, if the range of sampling times is wide enough to  
59 allow accumulation of substantial genetic differences between the first and last sampling  
60 event (Drummond et al. 2003). This is often the case for rapidly evolving viruses (Faria  
61 et al. 2014; Smith et al. 2009; Gire et al. 2014), and is starting to become possible to some  
62 degree for other organisms thanks to advances in ancient DNA (Orlando et al. 2014) and  
63 protein sequencing (Welker et al. 2015) technology. However, for macroevolutionary studies  
64 aiming to estimate divergence times on the order of tens or hundreds of million years, other  
65 sources of calibration information are required. Commonly, the age of the oldest known  
66 fossil of a given clade is then used to calibrate the age of this clade, an approach often  
67 referred to as “node dating” (Ronquist et al. 2012; Grimm et al. 2015). However, due to  
68 the incompleteness of the fossil record, clade origin will almost always predate the  
69 preservation of its oldest known fossil. As a result, fossils can provide absolute minimum  
70 clade ages, but are usually less informative regarding the maximum ages of clades (Benton

71 and Donoghue 2007). In a Bayesian framework for phylogenetic time calibration, the  
72 uncertainty regarding clade ages can be accommodated by the specification of “calibration  
73 densities” (also referred to as “node age priors”) with a hard lower bound set to the age of  
74 the earliest fossil record and a soft upper bound as provided by exponential, lognormal, or  
75 gamma distributions. Unfortunately, the optimal parameterization of these distributions is  
76 usually unknown but has been shown to have a strong influence on the resulting age  
77 estimates (Ho and Phillips 2009). In addition, the effect of inaccurate calibration densities  
78 can only partially be corrected with larger molecular datasets (Yang and Rannala 2006).

79 Other shortcomings of node dating have been identified. As described by Heled and  
80 Drummond (2012), calibration densities interact with each other and with the tree prior to  
81 produce marginal prior distributions of node ages that may differ substantially and in  
82 unpredictable ways from the specified calibration density. The application of  
83 recently-introduced calibrated tree priors can compensate for this effect, but becomes  
84 computationally expensive when more than a handful of calibrations are used in the  
85 analysis (Heled and Drummond 2015). Node dating has also been criticized for ignoring  
86 most of the information from the fossil record, as only the oldest known fossils of each  
87 clade are used to define calibration densities (Ronquist et al. 2012, but see Marshall 2008;  
88 Claramunt and Cracraft 2015). Furthermore, node dating relies on the correct taxonomic  
89 assignment of fossils to clades, and may produce misleading age estimates when fossils are  
90 misplaced on the phylogeny (Marshall 2008; Ho and Phillips 2009; Forest 2009).

91 As alternatives to node dating, two approaches have recently been developed. In  
92 “total-evidence” dating, fossils are not explicitly assigned to any clades, but are instead  
93 included as terminal taxa. The position of these tips is determined as part of the  
94 phylogenetic analysis, based on morphological character data that are required for all  
95 included fossils and at least some of the extant taxa (Pyron 2011; Ronquist et al. 2012).  
96 Branch lengths, and thus divergence times between extant and extinct species are inferred

97 under the assumption of a “morphological clock”, usually based on the Mk model of Lewis  
98 (2001) (Pyron 2011; Beck and Lee 2014; Arcila et al. 2015). The total-evidence approach is  
99 conceptually appealing as it is able to account for uncertainty in the phylogenetic position  
100 of fossils, and allows a more complete representation of the fossil record than node dating.  
101 However, this approach has been found to result in particularly ancient age estimates and  
102 long “ghost lineages” when applied to empirical data sets, leading some authors to question  
103 the suitability of morphological clocks for phylogenetic time calibration (Beck and Lee  
104 2014; Arcila et al. 2015; O’Reilly et al. 2015). The developments of more realistic sampling  
105 schemes (Höhna et al. 2011) and advanced models of morphological character evolution  
106 (Wright et al. 2016) are likely to improve age estimates obtained with total-evidence  
107 dating, but have so far been applied only rarely (Klopfstein et al. 2015; Zhang et al. 2016).  
108 Importantly, due to the requirement of a morphological character matrix, total-evidence  
109 dating is limited to groups that share sufficient numbers of homologous characters (Grimm  
110 et al. 2015) so that its application is practically not feasible for higher-level phylogenies  
111 combining very disparate taxa from different taxonomic orders or classes.

112 The “fossilized birth-death (FBD) process” (Stadler 2010) provides an elegant  
113 framework in which fossils are used as terminal taxa and thus more than the oldest fossil  
114 can be used for each clade. In this model, fossils as well as extant taxa are considered as  
115 the outcome of a common process based on the four parameters speciation rate  $\lambda$ ,  
116 extinction rate  $\mu$ , proportion of sampled extant taxa  $\rho$ , and the fossil “sampling rate”  $\psi$ . It  
117 is assumed that the fossils that are ultimately sampled and included in the study have been  
118 preserved along branches of the complete species tree following a constant-rate Poisson  
119 process. Unlike in total-evidence dating, a morphological character matrix is not required  
120 to place fossil taxa in the phylogeny (but can be used for this purpose; Gavryushkina et al.  
121 2015; Zhang et al. 2016). Instead, fossils are assigned to clades through the specification of  
122 topological constraints. The FBD process was first implemented in the Bayesian

123 divergence-time estimation program DPPDIV (Heath et al. 2014), which requires the  
124 specification of a fixed tree topology, point estimates of fossil ages, and constant rates of  
125 diversification and sampling throughout the tree. All these limitations have subsequently  
126 been overcome in the implementations of the FBD process as a tree prior in the “Sampled  
127 Ancestors” package for BEAST (Gavryushkina et al. 2014; Bouckaert et al. 2014) and in  
128 the software MrBayes (Huelsenbeck and Ronquist 2001; Ronquist and Huelsenbeck 2003;  
129 Zhang et al. 2016). These implementations allow one to specify priors on fossil ages to  
130 account for the often large uncertainties associated with them as well as the specification of  
131 time intervals within which rates are assumed constant, but between which they are free to  
132 vary. However, a limitation that remains also in newer FBD implementations is the  
133 assumption that all clades existing in a given time interval are subject to the same rates of  
134 diversification and fossil sampling. Especially in higher-level phylogenies, this assumption  
135 is unlikely to be met, as substantial clade-specific differences in these rates have been  
136 identified in many groups (Foote and Sepkoski 1999; Alfaro et al. 2009; Jetz et al. 2012;  
137 also see Supplementary Table S1), suggesting that time estimates obtained on the basis of  
138 this assumption may be misleading. The FBD model further assumes that the fossils  
139 included in the analysis represent either the complete set or a random sample of the known  
140 fossil record of a clade. However, the use of a complete or randomly sampled representation  
141 of the fossil record may be impractical with higher-level phylogenies due to the enormous  
142 number of fossil occurrences known for many higher taxa, e.g. for mammals (> 90 000),  
143 birds (> 5 000), and insects (> 40 000; [www.paleobiodb.org](http://www.paleobiodb.org)). Presumably as a consequence  
144 of these difficulties, node dating has remained popular despite its drawbacks, and was  
145 applied in all recent phylogenomic time-tree analyses of groups above the order level (dos  
146 Reis et al. 2015; Prum et al. 2015; Fernández et al. 2016).

147         Here, we develop a new approach for Bayesian phylogenetic divergence-time  
148 estimation that is related to node dating, but infers the optimal shape of calibration

149 densities from a combination of the first fossil occurrence age of a given clade and  
150 independently assessed estimates of the sampling rate and the diversification rates. This  
151 approach therefore overcomes a major problem of node dating, the fact that calibration  
152 densities are often chosen arbitrarily despite their strong influence on the resulting age  
153 estimates (Heath et al. 2014). In contrast to node dating, calibration densities in our  
154 approach are not directly applied to node ages, but to the age of origin of clades, and as a  
155 consequence, knowledge about the sister groups of calibrated clades is not required. Our  
156 approach is suitable for time calibration of higher-level phylogenies combining groups with  
157 different sampling characteristics, as the sampling rate can be specified independently for  
158 each clade. We have implemented our method in a new package for BEAST called  
159 “CladeAge”, and we will refer to calibration densities obtained with it as “CladeAge  
160 calibration densities” throughout the paper. Using a wide range of simulations, we assess  
161 the optimal scheme by which to select clades for calibration, and we show that the  
162 application of CladeAge calibration densities can result in age estimates comparable or  
163 better than those produced with the FBD model if the input rate estimates are correctly  
164 specified and only the oldest fossil of each clade is used for calibration.

165         We use our new approach together with a large and partially new molecular data  
166 set for 1187 teleost fishes to address the long-standing question whether freshwater cichlid  
167 fishes from India, Madagascar, Africa and the Neotropics diverged before or after  
168 continental separation (Chakrabarty 2004; Sparks and Smith 2005; Genner et al. 2007;  
169 Azuma et al. 2008; Friedman et al. 2013; McMahan et al. 2013). By rigorous examination  
170 of the paleontological literature, we identify the earliest fossil records for 147 out of 362  
171 (41%) well-characterized clades in our phylogeny. Our results strongly support divergence  
172 of freshwater fishes long after continental separation, implying multiple marine dispersal  
173 events not only in cichlid fishes but also in other freshwater groups included as outgroups  
174 in our phylogeny.

# CLADEAGE CALIBRATION DENSITIES

## *Calculating CladeAge Calibration Densities*

Here, our goal is to design calibration densities that reproduce the probability density for a clade originating at time  $t_o$ , given the age of its oldest fossil  $t_f$ . To estimate this probability density, we assume that the probability density of a clade being  $t$  time units older than its oldest fossil is identical to the probability density  $f_s(t)$  of the oldest fossil being  $t$  time units younger than the clade origin. This is equivalent to assuming a uniform prior probability distribution for the age of the clade, which is justified for calibration densities, as these probability densities will be multiplied with a (non-uniform) tree prior at a later stage, during the divergence time analysis. Thus, any non-uniform prior assumptions about the clade origin can be incorporated via the tree prior. We further assume that speciation, extinction, and fossil sampling are all homogeneous Poisson processes with rates  $\lambda$ ,  $\mu$ , and  $\psi$ , respectively.

For a single lineage that does not speciate or go extinct, the probability to remain unsampled until time  $t_1$  is  $p_u(t_1) = e^{-\psi t_1}$ , while the probability of being sampled at least once during the same period is  $p_s(t_1) = 1 - e^{-\psi t_1}$ . Thus the probability of not being sampled before time  $t_1$ , but then being sampled before time  $t_2$  (with  $t_2 > t_1$ , i.e., time  $t_2$  occurs after  $t_1$ ) is

$$\begin{aligned} p_{u,s}(t_1, t_2) &= e^{-\psi t_1} * (1 - e^{-\psi(t_2 - t_1)}) \\ &= -e^{-\psi t_2} + e^{-\psi t_1}. \end{aligned} \tag{1}$$

The probability density for the clade being sampled for the first time exactly at time  $t_1$  is then



$$\begin{aligned}
f_s(t_1) &= \lim_{t_2 \rightarrow t_1} \frac{-e^{-\psi t_2} + e^{-\psi t_1}}{t_2 - t_1} \\
&= - \lim_{t_2 \rightarrow t_1} \frac{e^{-\psi t_2} - e^{-\psi t_1}}{t_2 - t_1} \\
&= \psi e^{-\psi t_1}.
\end{aligned} \tag{2}$$

195            If we now allow for the possibility that the lineage has diversified into  $N$  species  
196 extant at time  $t_1$ , then the probability of the clade not being sampled before time  $t_1$  is  
197  $p_u(t_1) = e^{-\psi S(t_1)}$ , where  $S(t_1)$  is the sum of lineage durations between clade origin and time  
198  $t_1$  (Foote et al. 1999). The probability of no lineage being sampled before time  $t_1$ , but at  
199 least one lineage being sampled before time  $t_2$  is then

$$\begin{aligned}
p_{u,s}(t_1, t_2) &= e^{-\psi S(t_1)} * (1 - e^{-\psi(S(t_2) - S(t_1))}) \\
&= -e^{-\psi S(t_2)} + e^{-\psi S(t_1)}.
\end{aligned} \tag{3}$$

200            In this case, the probability density for the clade being sampled for the first time  
201 exactly at  $t_1$  is

$$\begin{aligned}
f_s(t_1) &= \lim_{t_2 \rightarrow t_1} \frac{-e^{-\psi S(t_2)} + e^{-\psi S(t_1)}}{t_2 - t_1} \\
&= - \lim_{t_2 \rightarrow t_1} \frac{e^{-\psi S(t_2)} - e^{-\psi S(t_1)}}{t_2 - t_1} \\
&= \psi e^{-\psi S(t_1)} * S'(t_1),
\end{aligned} \tag{4}$$

202            where the first derivative of  $S(t)$  at time  $t_1$  is

$$S'(t_1) = \lim_{t_2 \rightarrow t_1} \frac{S(t_2) - S(t_1)}{t_2 - t_1} \quad (5)$$

203 By ignoring the possibility of speciation or extinction between  $t_1$  and  $t_2$  (which is  
 204 justified at the limit  $t_2 \rightarrow t_1$ ), we get  $S(t_2) = S(t_1) + N * (t_2 - t_1)$  and thus  $S'(t_1) = N$ ,  
 205 which gives us

$$f_s(t_1) = \psi N e^{-\psi S(t_1)}. \quad (6)$$

206 If we now take into account the stochastic nature of  $S$  as a variable resulting from a  
 207 birth-death process with parameters  $\lambda$  and  $\mu$ , we have to rewrite Equation 6 as

$$f_s(t_1) = \mathbb{E}[\psi N e^{-\psi S(t_1)} | N \geq 1] \quad (7)$$

208 where we condition on the survival of at least one species at time  $t_1$ , which is  
 209 necessary to allow sampling at this time.

210 Unfortunately, we can not solve  $f_s(t_1)$  analytically. To approximate  $f_s(t_1)$ ,  
 211 CladeAge generates 10 000 birth-death trees based on estimates of the speciation rate  $\lambda$   
 212 and the extinction rate  $\mu$ , infers  $S(t_1)$  in each of these trees as the sum of all branch  
 213 lengths between clade origin and  $t_1$ , and calculates  $\psi N e^{-\psi S(t_1)}$  if the birth-death process  
 214 resulted in  $N \geq 1$ . According to the law of large numbers, the mean of a large sample  
 215 converges to its expected value, therefore the probability density  $f_s(t_1)$  can be  
 216 approximated by the mean of all values calculated for  $\psi N e^{-\psi S(t_1)}$ . This process is repeated  
 217 for 100 time points evenly spaced between 0 and a maximum time  $t_{max}$ , which is  
 218 predetermined so that the probability density for the clade being first sampled at this time

219  $(f_s(t_{max}))$  is negligible compared to the probability density for the clade being sampled at  
220 the very start of the process ( $f_s(0)$ ) (more specifically, the approximations  $N(t) = e^{(\lambda-\mu)*t}$   
221 and  $S(t) = \int_0^t N(t) dt$  are used to find a solution for  $f_s(t_{max}) = 0.001 * f_s(t_0)$ ). The  
222 probability density  $f_s(t)$  for times  $t$  in between two of the 100 time points is estimated  
223 through interpolation from the probability densities of the two neighbouring time points,  
224 using linear regression. For all times larger than  $t_{max}$ , probability densities are  
225 approximated by a scaled exponential distribution that is calculated on the basis of the two  
226 largest time points and their respective probability densities  $f_s(t)$ . Finally, all estimates of  
227 probability densities are scaled so that the total probability mass becomes 1.

228         The calculation of calibration densities, as described above, requires estimates of the  
229 fossil sampling rate, as well as of the speciation and extinction rates, which can be  
230 obtained externally, from the fossil record alone (Silvestro et al. 2014; Starrfelt and Liow  
231 2016), or from a combination of fossil and phylogenetic information (Alfaro et al. 2009;  
232 Stadler 2011; Rabosky 2014). As diversification is commonly parameterized as “net  
233 diversification” ( $\lambda - \mu$ ) and “turnover” ( $\mu/\lambda$ ), and researchers often have greater confidence  
234 in estimates of net diversification and turnover than in those for speciation and extinction  
235 rates (Beaulieu and Donoghue 2013), our method accepts input in these units, and  
236 calculates  $\lambda$  and  $\mu$  from it. Uncertainty in the three parameters net diversification,  
237 turnover, and sampling rate can be expressed by specifying minimum and maximum values  
238 and is accounted for by randomly drawing from the specified ranges, for each of the 10 000  
239 birth-death trees generated to estimate estimated probability density  $f_s$ . Examples  
240 demonstrating the shape of CladeAge calibration densities, based on exactly known (A) or  
241 uncertain ages of the first fossil record (B), are shown in Figure 1.

242         *Calibration Schemes for the Use of CladeAge Calibration Densities in*  
243                 *Phylogenetic Divergence-Time Estimation*

244 Under the assumption of constant rates of diversification and sampling as well as a  
245 uniform prior probability for node ages, CladeAge calibration densities approximate the  
246 probability density for the age of a clade, given the age of the oldest fossil record of this  
247 clade. These probability distributions are therefore suitable as constraints on clade ages in  
248 Bayesian divergence-time estimation. However, in practice, it may not always be clear  
249 which clades should be used for time calibration: If a fossil represents the earliest record of  
250 not only one clade, but of multiple nested clades, CladeAge calibration densities could be  
251 used to constrain the age of origin of all these clades (we refer to this as “scheme A”), only  
252 of the most inclusive of these clades (“scheme B”), or only of the least inclusive clade  
253 (“scheme C”). As scheme B would allow one or more of the clades to appear younger than  
254 the fossil itself, it seems reasonable to specify, in addition to the CladeAge calibration  
255 density for the most inclusive clade, the fossil age as a strict minimum age for the least  
256 inclusive clade when using this scheme. Furthermore, if two sister clades both possess a  
257 fossil record, these fossils could be used to constrain the ages of both of the two clades.  
258 However, as the ages of the two clades are necessarily linked by their simultaneous  
259 divergence, two time constraints would effectively be placed on one and the same node.  
260 Instead, it may seem more intuitive to use only the older of the two fossils for time  
261 calibration and disregard the younger fossil (“scheme D”). However, in contrast to node  
262 dating, where maximally one calibration density is placed on each node, the model used to  
263 calculate CladeAge calibration densities considers each clade individually, and could thus  
264 be biased if the selection of clades for calibration is based on information about their sister  
265 clade. Figure 2a illustrates the four different calibration schemes.

266 As CladeAge calibration densities approximate the probability densities of clade  
267 ages conditional on the age of the first fossil record of this clade, they are also expected to  
268 approximate frequency distributions of observed waiting times between the origin of a  
269 clade and the appearance of the first fossil record of this clade in a sufficiently large sample

270 of simulated phylogenetic trees. Since these waiting times can be sampled according to the  
271 above four schemes, we can determine the optimal calibration scheme by comparison of  
272 waiting time frequency distributions with CladeAge calibration densities. We simulated  
273 three times 10 000 pure-birth phylogenies with a speciation rate  $\lambda = 0.04$  and a root age  
274  $t_{root}$  randomly drawn from a uniform distribution between 20 and 200 time units,  
275 conditioned on the survival of exactly 100 extant species. Assuming a Poisson process of  
276 fossil sampling, we added simulated fossil records to the branches of each of these trees,  
277 with three different sampling rates  $\psi = 0.1, 0.03, 0.01$ . Applying the above four calibration  
278 schemes (A-D) independently, we recorded waiting times between a clade's origin and the  
279 age of its oldest fossil in each simulated phylogeny.

280         Waiting time frequency distributions recorded from relatively young clades can be  
281 biased by the fact that only those waiting times shorter than the clade age can be recorded  
282 (otherwise the clade did not preserve at all). To assess the degree of this effect, we repeated  
283 this analysis, counting only waiting times for clades with a time of clade origin  $t_o$  above  
284 one out of four thresholds:  $t_o \geq 0$  (all clades included),  $t_o \geq 0.5 \times t_{root}$ ,  $t_o \geq 0.9 \times t_{root}$ , and  
285  $t_o = 1 \times t_{root}$  (including only the two clades descending from the root, per simulated  
286 phylogeny). With the strictest clade age threshold of  $t_o = 1 \times t_{root}$ , the same two waiting  
287 times per phylogeny are recorded with schemes A and B if both clades descending from the  
288 root have produced fossils. This is because the root node represents the oldest node that  
289 can be constrained with fossils in these clades, and thus waiting times between the root and  
290 these fossils are recorded with both schemes A and B. If further divergence events occurred  
291 between the root and the fossil, the root does not represent the youngest node that can be  
292 constrained with the fossil, and thus, the waiting time between the root and the fossil are  
293 not recorded with schemes C and D (see Fig. 2a). Differences between schemes A and B  
294 become apparent with less strict clade age thresholds, when also clades are included that  
295 do not represent the oldest possible clade to be constrained with a given fossil.

296 Figure 2b shows comparisons between waiting time frequency distributions and  
297 CladeAge calibration densities for a clade age threshold of  $t_o \geq 0.9 \times t_{root}$ , which is  
298 sufficiently young to show differences between all schemes, but still old enough to be  
299 affected only minimally by the bias described above. Comparisons for all other tested clade  
300 age thresholds are shown in Supplementary Figure S1. Taken together, these results show  
301 that waiting time frequency distributions deviate from the respective CladeAge  
302 distribution in most comparisons, and the degree of disagreement depends on sampling rate  
303  $\psi$ , on the clade age threshold, and on the applied scheme (A-D). However, for all but the  
304 youngest clade age thresholds, scheme A produces a frequency distribution that is virtually  
305 identical in shape to the distribution of CladeAge calibration densities. This suggests that  
306 when CladeAge calibration densities are used for time calibration, they should strictly be  
307 applied to constrain all clades for which a given fossil represents the first occurrence, even  
308 if the same fossil is used to constrain multiple nodes, and even if more than one constraint  
309 is placed on one and the same node.

## 310 TESTING CLADEAGE CALIBRATION DENSITIES WITH 311 SIMULATED PHYLOGENIES

312 To more extensively compare the performance of the four different calibration  
313 schemes A to D, we simulated phylogenetic data sets including fossil records and sequence  
314 alignments, and used CladeAge calibration densities to estimate clade ages in BEAST  
315 v.2.1.3. For comparison, we also used the same generated data sets to estimate clade ages  
316 with the FBD model implemented in the Sampled Ancestors (Gavryushkina et al. 2014)  
317 package for BEAST.

### 318 *Generation of Data Sets*

319 Phylogenetic data sets of trees and fossil records were generated as described above  
320 with sampling rates  $\psi = 0.1, 0.03, 0.01$ , a root age between 20 and 200 time units, and a  
321 net diversification  $\lambda - \mu$  of 0.04, however, species turnover was now modeled with rate  
322  $\mu/\lambda = 0.5$  (thus using  $\lambda = 0.08$  and  $\mu = 0.04$ ). If the time units used in these simulations  
323 are considered to be million years, the sampling and diversification rates used here are  
324 comparable to those found in empirical data sets (Jetz et al. 2012; Stadler and Bokma  
325 2013; Rabosky et al. 2013; Supplementary Table S1). In separate sets of simulations,  
326 branch-specific substitution rates were modeled either with an uncorrelated molecular clock  
327 (Drummond et al. 2006), or with an autocorrelated molecular clock that accounts for the  
328 heritability of factors influencing rate variation (such as body mass, longevity, and  
329 generation time; Nabholz et al. 2008; Amster and Sella 2016) and may therefore model rate  
330 evolution more realistically than the uncorrelated molecular clock (Lepage et al. 2007). For  
331 both types of branch rate variation, we used a mean rate of  $4 \times 10^{-3}$  substitutions per site  
332 per time unit and a variance parameter of  $1.6 \times 10^{-5}$ . Branch-rate autocorrelation was  
333 simulated with the Cox-Ingersoll-Ross (CIR) process as described by Lepage et al. (2006),  
334 using a decorrelation time of 100 time units. The branch lengths and substitution rates  
335 were used to simulate sequence evolution of 3000 nucleotides according to the unrestricted  
336 empirical codon model of Kosiol et al. (2007). For each of two clock models and each of the  
337 three sampling rates, we generated 50 replicate data sets. An example of a data set  
338 simulated with these settings is illustrated in Supplementary Figure S2.

### 339 *Phylogenetic Divergence-Time Estimation*

340 For each of the replicate data sets, the simulated phylogenetic trees were  
341 reconstructed, and for each clade in each reconstructed phylogeny, the oldest fossil  
342 occurrence was identified. CladeAge calibration densities were calculated for these fossils  
343 based on the parameters used in simulations ( $\lambda = 0.08, \mu = 0.04$ , and  $\psi = 0.1, 0.03, 0.01$ ),

344 and used to constrain node ages according to calibration schemes A to D in  
345 divergence-time estimation with BEAST. To exclude the possibility that clades appear  
346 younger than their fossils in scheme B (see Figure 2a), additional uniform calibration  
347 densities were used in this scheme for the ages of all clades with fossils. These uniform  
348 densities were specified using the fossil age as a hard lower boundary and an unrealistically  
349 high upper boundary (arbitrarily placed at 1000 time units) to avoid improper prior  
350 distributions. All sequence alignments were divided into three partitions according to  
351 codon position, and for each partition, we used the reversible-jump-based substitution  
352 model of Bouckaert et al. (2013) with four gamma-distributed rate categories. For  
353 divergence-time estimation with all simulated data sets, we used the lognormal relaxed  
354 molecular clock (Drummond et al. 2006). To account for extinction in the diversification  
355 process, we applied the birth-death tree prior of Gernhard (2008) with uninformative prior  
356 distributions for the birth rate and the relative death rate. We used the reconstructed  
357 simulated tree as a starting tree in all analyses, and fixed the tree topology by disallowing  
358 all topological changes. For each analysis, 50 million Markov-chain Monte Carlo (MCMC)  
359 steps were carried out, which was always sufficient for convergence.

360 For the analysis of the same data sets with the FBD model implemented in the  
361 Sampled Ancestors package for BEAST, we used settings as described above, except that  
362 between 100 and 400 million MCMC steps were required for convergence. For  
363 comparability with age estimates based on CladeAge calibration densities, we again used  
364 only the oldest fossil for each clade. While this reduction of the simulated fossil record to  
365 the oldest fossil of each clade represents a violation of the assumptions of the FBD model,  
366 we were interested in the performance of the FBD model in this scenario, as in practice the  
367 information about the oldest fossil record of a clade is often easier to obtain and implement  
368 in the analysis (see Discussion). The values of diversification rates were fixed to those used  
369 to generate the data set, however, the sampling proportion was either fixed according to



370 the sampling rate used in simulations or allowed to be estimated in separate analysis  
371 replicates. We also fixed the tree topology of all extant species, while at the same time  
372 allowing fossil taxa to attach anywhere within the clade (including its stem lineage) to  
373 which they were assigned. This was done by using instances of “CladeConstraint”, a new  
374 type of topological constraint for BEAST introduced as part of the Sampled Ancestors  
375 package (Gavryushkina et al. 2014), with which ingroups and outgroups can be defined for  
376 a given clade, and taxa not listed in either of these groups are free to appear in either of  
377 them. For each clade, we specified CladeConstraints that place all extant taxa and fossils  
378 of this clade within the ingroup and all other extant taxa in the outgroup, thus allowing  
379 fossils from parent clades to appear outside or within this clade. As the starting tree, we  
380 used the reconstructed simulated tree but reattached each clade’s oldest fossil (provided  
381 that it had any) to its stem lineage with an additional branch.

382 To test the robustness of our approach to parameter misspecification, we repeated  
383 all analyses in which calibration scheme A was used with CladeAge calibration densities  
384 calculated on the basis of net diversification rates and sampling rates that were different  
385 from those used to generate data sets. In these analyses, the net diversification rate used  
386 for inference was chosen as either 25% larger or smaller than the true net diversification  
387 rate, or the sampling rate was set to either 50% larger or smaller than the true sampling  
388 rate used for simulations. We also applied the same misspecified rates for net  
389 diversification and sampling in separate analyses with the FBD model to allow a  
390 comparison of the robustnesses of the CladeAge and FBD models. BEAST input files used  
391 for the analysis of simulated datasets are provided as Supplementary Data S1.

### 392 *Results with Simulated Phylogenies*

393 Our simulations produced phylogenetic trees with root heights between 52.2 and  
394 163.8 time units, with a median height of 84.8 time units. Mean branch rates per tree were

395 between  $2.2 \times 10^{-3}$  and  $6.6 \times 10^{-3}$  (median  $3.9 \times 10^{-3}$ ) substitutions per time unit with  
396 branch rate variances between  $7.8 \times 10^{-7}$  and  $4.4 \times 10^{-5}$  (median  $7.1 \times 10^{-6}$ ), resulting in  
397 5280 to 15090 (median 8289) nucleotide substitutions. The sequence alignments contained  
398 between 2217 and 2836 (median 2568) variable sites and between 1724 and 2646 (median  
399 2231) parsimony-informative sites, out of a total of 3000 sites per alignment. Simulated  
400 fossil records consisted of 165 to 380 (median 240.5) fossils when generated with a sampling  
401 rate of  $\psi = 0.1$ , 40 to 123 (median 74.5) fossils with  $\psi = 0.03$ , and 10 and 49 (median 24)  
402 fossils when a sampling rate of  $\psi = 0.01$  was applied (see Supplementary Figure S2 for an  
403 illustration). Discarding fossils that did not represent the oldest fossil of any clade left  
404 between 75 and 113 (median 94) fossils when the sampling rate was  $\psi = 0.1$ , 30 to 71  
405 (median 47) fossils with  $\psi = 0.03$ , and 9 to 33 (median 19.5) fossils with  $\psi = 0.01$ . Figure  
406 3a shows the mean number of fossil constraints in 50 simulated data sets, per bin of 20 time  
407 units. The number of fossils available as time constraints decreases with bin age, a direct  
408 result of the fact that younger time bins contain an overall larger sum of lineage durations.

409 Comparisons of estimated and true node ages are shown in Figure 3b-c, for all  
410 analyses of data sets generated with the intermediate sampling rate of  $\psi = 0.03$  and the  
411 uncorrelated clock model (results obtained with  $\psi = 0.1$  or  $\psi = 0.01$ , or with the  
412 autocorrelated clock model are provided in Supplementary Figure S3, and results of  
413 robustness tests with misspecified rates are shown in Supplementary Figure S4). The  
414 difference between results based on MCMC sampling from the prior only (Fig. 3b) and  
415 results based on the posterior (Fig. 3c) is most pronounced for young clades where 95%  
416 highest posterior density (HPD) intervals (indicated with gray bars in Fig. 3b-c) are much  
417 wider when the MCMC sampled from the prior only. This suggests that in combination  
418 with a relaxed clock model, sequence data is most informative to determine the age of  
419 young nodes, but that the age estimates of older nodes are primarily determined by the  
420 specified prior probabilities.

421           Following Heath et al. (2014) and Gavryushkina et al. (2014), we describe the age  
422 estimates for simulated phylogenies with two summary statistics, the mean width of 95%  
423 HPD intervals and the percentage of 95% HPD intervals that include the true node age.  
424 Shorter 95% HPD intervals indicate greater precision, and the percentage of 95% HPD  
425 intervals that include the true node age serves to assess the accuracy of age estimates. If  
426 the model used to generate the data is identical to that assumed for divergence-time  
427 estimation, and if MCMC sampling has completely converged, 95% of the 95% HPD  
428 intervals are expected to include the true node age. For CladeAge analyses of data sets  
429 generated with uncorrelated branch rates, a nearly identical model was used for simulation  
430 and inference, and the resulting percentage of 95% HPD intervals containing the true node  
431 age can therefore serve as an indicator of the optimal calibration scheme to be used with  
432 CladeAge. In contrast, the model used in analyses with the FBD differs to a greater extent  
433 from the model used to generate data sets, as the FBD model assumes that all, or a  
434 randomly sampled set of fossils of a clade are used for calibration, whereas our data sets  
435 were reduced to contain only the oldest fossils of each clade. Thus, for FBD analyses of  
436 data sets generated with uncorrelated branch rates, the two summary statistics allow to  
437 assess the robustness of the FBD model to a violation of the assumed fossil record  
438 representation. In addition, the robustness of both CladeAge and FBD analyses to further  
439 model violation is indicated by results for data sets generated with autocorrelated branch  
440 rates, and by inferences based on misspecified rates of net diversification and sampling.

441           For all analyses in which rates were either correctly specified or allowed to be  
442 estimated, the two summary statistics are listed in Table 1, and illustrated in bins of 20  
443 time units in Figure 3d-e for data sets generated with the uncorrelated clock model  
444 (detailed results for all analyses are given in Supplementary Tables S10-S12). For  
445 robustness tests with misspecified rates, the two summary statistics are listed in  
446 Supplementary Tables S13-S14.

447           Among the four calibration schemes A to D, scheme A produced the shortest 95%  
448 HPD intervals with data sets based on  $\psi = 0.1$  or  $\psi = 0.03$ , regardless of whether the  
449 MCMC was set to sample from the prior only, or from the posterior, and both with data  
450 sets generated with uncorrelated or autocorrelated branch rates. At the same time, the  
451 percentage of true node ages included in 95% HPD intervals obtained with scheme A is  
452 closer to the expected value of 95% than that of any other calibration scheme. In contrast,  
453 scheme B performed slightly better than scheme A for data sets with the lowest sampling  
454 rate  $\psi = 0.01$ , as indicated by shorter 95% HPD intervals and a greater percentage of true  
455 node ages included within them. However, when scheme B was used for the analysis of data  
456 sets generated with uncorrelated branch rates, the accuracy of age estimates decreased  
457 with node age, and for nodes with a true age between 80 and 100 time units, only 76.2% of  
458 the 95% HPD intervals contained their true age (Fig. 3e, Supplementary Table S11).

Table 1: Estimated node ages for simulated phylogenies, based on four CladeAge calibration schemes and the FBD model.

<b>Mean 95% HPD width:</b>							
Clock model	$\psi$	Scheme A	Scheme B	Scheme C	Scheme D	FBD (fixed $\psi$ )	FBD (est. $\psi$ )
prior only	0.1	9.34	10.30	11.55	14.07	11.84	20.01
prior only	0.03	17.42	18.45	21.84	24.27	18.11	22.30
prior only	0.01	25.28	24.13	33.26	35.72	22.03	23.66
uncorrelated	0.1	6.41	6.91	7.55	9.00	8.20	12.68
uncorrelated	0.03	10.63	11.06	12.91	14.30	11.90	14.39
uncorrelated	0.01	14.08	13.19	18.67	19.99	13.71	14.77
autocorrelated	0.1	4.56	4.82	5.10	6.03	5.75	8.66
autocorrelated	0.03	6.77	6.92	7.83	8.73	7.84	9.44
autocorrelated	0.01	8.49	8.00	11.39	12.63	8.76	9.61

  

<b>Percentage of 95% HPD intervals containing the true node age:</b>							
Clock model	$\psi$	Scheme A	Scheme B	Scheme C	Scheme D	FBD (fixed $\psi$ )	FBD (est. $\psi$ )
prior only	0.1	95.2	96.7	96.1	95.6	98.2	93.0
prior only	0.03	94.8	94.4	93.2	91.5	97.0	96.1
prior only	0.01	92.6	94.3	89.5	88.5	96.1	96.5
uncorrelated	0.1	94.9	95.4	95.3	93.8	95.8	83.7
uncorrelated	0.03	93.7	93.2	90.9	88.3	94.7	91.1
uncorrelated	0.01	90.5	92.4	83.9	82.0	94.1	94.0
autocorrelated	0.1	87.9	87.3	86.4	81.9	84.4	63.5
autocorrelated	0.03	76.8	75.7	69.3	62.7	72.2	63.8
autocorrelated	0.01	66.2	70.3	57.0	54.3	70.3	67.7

Notes: Divergence-time estimation was based on MCMC sampling from prior probabilities alone, or in combination with the likelihood of sequence data simulated with uncorrelated or autocorrelated branch rates. Fossil records were simulated with three different sampling rates  $\psi = 0.1, 0.03, 0.01$ . For the FBD model, results are shown for analyses in which the sampling rate  $\psi$  was either fixed or allowed to be estimated.

459 In all cases, MCMC sampling from the posterior decreased the mean width of 95%  
460 HPD intervals, compared to analyses using the prior probability alone. The percentage of  
461 95% HPD intervals containing the true node age remained comparable between analyses  
462 based on the prior probability alone (92.6-95.2% with scheme A) and analyses using the  
463 posterior (90.5-94.9% with scheme A) for data sets generated with uncorrelated branch  
464 rates. However, for data sets generated with autocorrelated branch rates, the percentage of  
465 95% HPD intervals containing the true node age decreased substantially when the  
466 posterior was used for MCMC sampling (66.2-87.9% with scheme A; Table 1).

467 Overall, FBD analyses with a fixed sampling rate produced very similar summary  
468 statistics to CladeAge analyses with scheme A (Table 1). As for CladeAge analyses, the  
469 percentage of 95% HPD intervals containing the true node age was lower with  
470 autocorrelated branch rates, and remained around 95% with uncorrelated branch rates or  
471 when MCMC sampling from the prior only. In seven out of nine comparisons, however, the  
472 95% HPD intervals were slightly wider when estimated with the FBD model than with  
473 CladeAge scheme A. The FBD model, used with a fixed sampling rate, also appeared  
474 somewhat less robust to the violation of the assumed clock model (i.e. with branch-rate  
475 autocorrelation), except when the lowest sampling rate  $\psi = 0.01$  was used for dataset  
476 generation.

477 In contrast, when the sampling rate was not fixed in FBD analyses, 95% HPD  
478 intervals remained similarly wide regardless of the true sampling rate used in dataset  
479 generation, and relatively small percentages of 95% HPD intervals contained the true node  
480 age in analyses using the posterior (Fig. 3c). A particularly low percentage of 95% HPD  
481 intervals (63.5-63.8%; Table 1) contained the true node ages in analyses of datasets  
482 generated with autocorrelated branch rates and high or intermediate sampling rates  
483 ( $\psi = 0.1$  or  $\psi = 0.03$ ). Low accuracy with the FBD model in which sampling rates were  
484 not fixed was mostly due to overestimation of intermediate node ages (Fig. 3e,

485 Supplementary Figure S3b-c). The overestimation of node ages in these analyses coincides  
486 with a substantial underestimation of the sampling rate itself (Supplementary Figure S5).  
487 In analyses using the prior alone, the sampling rate was on average estimated as only 30.4,  
488 51.4, and 73.9% of the true sampling rate, when the true sampling rate was  $\psi = 0.1$ , 0.03,  
489 or 0.01, respectively. Also when sampling from the posterior in analyses of datasets  
490 generated with uncorrelated or autocorrelated branch rates, these percentages remained  
491 nearly identical (Supplementary Figure S3).

492 Both the CladeAge model and the FBD model appeared mostly robust to  
493 misspecification of rate estimates. The specification of a net diversification rate that is  
494 either 25% larger or smaller than the net diversification used to generate data sets has very  
495 little effect on both the mean width of 95% HPD intervals and the percentage of 95% HPD  
496 intervals containining the true node age, and this is so for analyses with both CladeAge  
497 (using calibration scheme A) and the FBD model (Supplementary Tables S13-S14). The  
498 strongest effects of parameter misspecification were found when the sampling rate specified  
499 for inference was only 50% of the true sampling rate used in simulations. In this case, node  
500 ages tended to be overestimated (Supplementary Figure S4) and the percentage of 95%  
501 HPD intervals containining the true node age dropped to 74.3% when CladeAge was used  
502 to analyze data sets generated with a low true sampling rate of  $\psi = 0.01$  (Supplementary  
503 Table S13). The FBD model performed better than CladeAge in analyses of data sets  
504 generated with a low true samplinig rate, with 92.2% of the 95% HPD intervals  
505 containining the true node age. In contrast, when the true sampling rate was high  
506 ( $\psi = 0.1$ ) but misspecified as 50% too low in the inference, 95% HPD intervals resulting  
507 from analyses with CladeAge contained more true node ages (93.8% vs. 90.4%) and were  
508 less wide (8.60 vs. 10.73) than those produced by the FBD model (Supplementary Tables  
509 S13-S14).

510 APPLYING CLADEAGE CALIBRATION DENSITIES TO  
511 RESOLVE DIVERGENCE TIMES OF CICHLID FISHES

512 *Phylogeography of Cichlidae*

513 Fishes of the percomorph family Cichlidae are known for their extraordinary species  
514 richness, which includes the replicated adaptive radiations in the Great Lakes of East  
515 Africa (Salzburger et al. 2014). Three reciprocally monophyletic subfamilies occur in Africa  
516 and the Middle East (Pseudocrenilabrinae; see Supplementary Text S2), in South and  
517 Central America (Cichlinae), and on Madagascar (Ptychochrominae). In addition, the  
518 most ancestral subfamily Etroplinae consists of two genera, of which one occurs in  
519 Southern India and Sri Lanka and another is endemic to Madagascar (Sparks and Smith  
520 2004). As the distribution of cichlids is mostly limited to landmasses of the former  
521 supercontinent Gondwana, their biogeography is traditionally considered a product of  
522 Gondwanan vicariance (Chakrabarty 2004; Sparks and Smith 2005; Smith et al. 2008;  
523 Azuma et al. 2008). In this scenario, the divergence of African and South American  
524 cichlids must have occurred before or during the break-up of the two continents about 100  
525 Ma (Heine et al. 2013), and Indian and Malagassy cichlids must have separated before 85  
526 Ma (Ali and Aitchison 2008). Regardless of whether cichlids colonized Africa or South  
527 America first, this colonization should have occurred before 120 Ma, as Madagascar and  
528 India were separated by that time from both Africa and Antarctica, through which a  
529 connection to South America could have existed previously (Ali and Aitchison 2008; Ali  
530 and Krause 2011).

531 However, a Gondwanan history is not supported by the fossil record of Cichlidae.  
532 Their earliest record is provided by †*Mahengechromis* spp. from Tanzania (46-45 Ma)  
533 (Murray 2000a), followed by the first occurrences of neotropical cichlids in the Argentinian



534 Lumbrera Formation (Malabarba et al. 2006; Alano Perez et al. 2010; Malabarba et al.  
535 2010). The age of the fossils of the Lumbrera Formation is often cited as 48.6 Ma (e.g.  
536 Alano Perez et al. 2010); however, the basis of this precise age estimate is questionable (see  
537 Supplementary Text S2 and Friedman et al. 2013; Benton et al. 2015). The Lumbrera  
538 Formation has been assigned to the Casamayoran age (45.4-38.0 Ma) (Vucetich et al. 2007;  
539 del Papa et al. 2010), and the age of the fossils can be further constrained by a minimum of  
540 39.9 Ma based on radiometric dating (del Papa et al. 2010; this age was incorrectly  
541 specified as 33.9 Ma in Friedman et al. 2013). Thus, we here assume an age of 45.4-39.9 Ma  
542 for the cichlid fossils of the Lumbrera Formation.

543         Due to the lack of cichlid remains older than 46 Ma, long ghost lineages would need  
544 to be postulated to reconcile the biogeography of cichlid fishes with Gondwanan vicariance.  
545 On the other hand, trans-oceanic dispersal over hundreds or thousands of kilometers,  
546 followed by successful colonization of a new continent, appears extremely improbable, given  
547 that cichlids are found almost exclusively in freshwater. Whereas several cichlid species  
548 occur in brackish-water estuaries and some species are known to tolerate marine saltwater  
549 conditions (Myers 1949; Stickney 1986; Uchida et al. 2000), none have ever been observed  
550 in the open ocean, more than a few miles from the coast (Conkel 1993; Greenfield and  
551 Thomserson 1997). Thus, a long-standing debate has centered on the relative probabilities  
552 of the two alternative scenarios, Gondwanan vicariance or trans-oceanic dispersal (Vences  
553 et al. 2001; Murray 2001a; Chakrabarty 2004; Sparks and Smith 2005; Genner et al. 2007;  
554 Smith et al. 2008). However, arguments for both sides have mostly been verbal, and the  
555 probabilities of the long ghost lineages required for the Gondwanan vicariance scenario  
556 could not properly be quantified, as an objective basis has been lacking for the specification  
557 of calibration densities in previous divergence-time analyses (e.g. Azuma et al. 2008; but  
558 see Friedman et al. 2013). In contrast, CladeAge calibration densities are based on  
559 sampling rate estimates and thus directly account for probabilities of individual ghost

560 lineage durations. In combination with a large-scale molecular phylogeny including  
561 multiple cichlid and outgroup fossil constraints, the CladeAge method is therefore ideally  
562 suited to assess the most plausible phylogeographic scenario for cichlid fishes.

### 563 *A Multi-Marker Phylogeny of Teleost Fishes*

564 In order to time-calibrate cichlid divergences, we applied CladeAge calibration  
565 densities to a large-scale phylogeny of cichlid and outgroup taxa, including nearly 150 fossil  
566 constraints. As a first step, we compiled a molecular data set for 40 mitochondrial and  
567 nuclear markers, sequenced from 1187 species of the teleost Supercohort Clupeocephala  
568 (see Betancur-R et al. 2013). Of the species included in the data set, 578 were members of  
569 the family Cichlidae, 516 were members of other families of Cohort Euteleostomorpha,  
570 and 93 species were members of Cohort Otomorpha, the sister lineage of  
571 Euteleostomorpha, and were collectively used as an outgroup in our phylogenetic analysis.  
572 Out of a total of 11 050 sequences, 9970 were retrieved from the NCBI nucleotide database  
573 ([www.ncbi.nlm.nih.gov/nucleotide](http://www.ncbi.nlm.nih.gov/nucleotide)), 85 were obtained from annotated genomes of the  
574 Ensembl database (Cunningham et al. 2015), 5 mt-co1 sequences were downloaded from  
575 the Barcode of Life Data System (BOLD; Ratnasingham and Hebert 2007), and 328  
576 sequences were identified from other non-annotated genomic resources (Supplementary  
577 Tables S2-S7). In addition, 662 sequences of 19 markers were produced specifically for this  
578 study, including 26 mitochondrial genomes (see Supplementary Text S1 for sequencing  
579 protocols and Supplementary Tables S2 and S5 for accession numbers).

580 For each marker, sequences were aligned with MAFFT v.7.122b (Kato and  
581 Standley 2013), visually inspected, and poorly aligned regions were removed. Alignments  
582 were subsequently divided into primary data blocks according to codon position. In  
583 combination, the alignments included 35 817 sites with an overall proportion of  
584 undetermined characters of 82.84%. Assuming a general time-reversible model of sequence

585 evolution with gamma-distributed rate variation among sites (GTR+ $\Gamma$ ), the fit of  
586 partitioning schemes was assessed according to the Bayesian Information Criterion (BIC).  
587 The best-fitting partitioning scheme determined with the greedy algorithm implemented in  
588 PartitionFinder v.1.0.1 (Lanfear et al. 2012) combined primary data blocks into 30  
589 different partitions (Supplementary Table S8).

590 Maximum likelihood (ML) phylogenetic tree search was conducted with RAxML  
591 v.7.3.1 (Stamatakis 2006; Pfeiffer and Stamatakis 2010), applying unlinked “GTRCAT”  
592 models of sequence evolution for each of the 30 partitions. Topological node support was  
593 evaluated with RAxML’s rapid bootstrap analysis (option “-f a”) and the “autoMRE”  
594 automatic stopping criterion (Stamatakis et al. 2008). Based on the ML phylogeny, we  
595 identified 455 clades that were potentially suitable for time calibration, as they were  
596 supported by high bootstrap values in our study ( $\geq 93\%$  with only 6 exceptions) and  
597 corroborated by previously published molecular phylogenetic analyses and morphological  
598 synapomorphies (Supplementary Figure S6 and Supplementary Text S2). Of the 455  
599 clades, 362 were mutually exclusive and in their sum represented nearly the entire species  
600 richness of Clupeocephala ( $> 99.5\%$ ; Supplementary Table S9). This is important in  
601 analyses with CladeAge calibration densities, as it ensures that the sister groups of clades  
602 used for time calibration are present in the phylogeny, even if their identity is not known  
603 prior to the phylogenetic analysis. If sister groups of clades with fossils were instead  
604 missing from the taxon set, the CladeAge calibration density based on their fossil record  
605 would not, as intended, apply to the age of origin of these clades, but only to the ages of  
606 origin of more inclusive clades, potentially leading to underestimation of divergence ages.  
607 To illustrate this point, imagine that clade A was missing in Figure 2a, then a CladeAge  
608 calibration density based on fossil  $F_2$  could not be used to calibrate the age of origin of  
609 clade B ( $O_2$ ), but only those of clades C and D ( $O_3$  and  $O_4$ ).

## CladeAge Model Parameter Estimation

610

611 CladeAge calibration densities are calculated based on estimates of rates of  
612 sampling ( $\psi$ ), net diversification ( $\lambda - \mu$ ), and turnover ( $\mu/\lambda$ ). In order to use CladeAge  
613 calibration densities for the time calibration of teleost divergences, we obtained estimates  
614 for these three parameters from previous studies. Net diversification and turnover rates of  
615 teleost fishes were estimated by Santini et al. (2009) as 0.041-0.081 per lineage per million  
616 years ( $L^{-1}\text{myr}^{-1}$ ) and 0.0011-0.37  $L^{-1}\text{myr}^{-1}$ , respectively. These estimates are comparable  
617 to those of a more recent analysis by Rabosky et al. (2013), who estimated a mean net  
618 diversification rate of 0.098  $L^{-1}\text{myr}^{-1}$  and a mean turnover rate of 0.284  $L^{-1}\text{myr}^{-1}$  using a  
619 Bayesian model of diversification with rate shifts. We here apply the slightly lower  
620 diversification rate estimates of Santini et al. (2009) (net diversification rate: 0.041-0.081  
621  $L^{-1}\text{myr}^{-1}$ ; turnover rate: 0.0011-0.37  $L^{-1}\text{myr}^{-1}$ ) to calculate CladeAge calibration  
622 densities and note that their distributions will tend to be wider, and thus older, than  
623 distributions calculated with the rate estimates of Rabosky et al. (2013).

624 Sampling probabilities have been estimated from the fossil record for a variety of  
625 groups and with a wide range of methods. For bony fishes (Osteichthyes) including  
626 Clupeocephala, an estimate of the sampling probability was calculated by Foote and  
627 Sepkoski (1999) from the frequency ratio  $f_2^2/(f_1f_3)$ , where  $f_1$ ,  $f_2$ , and  $f_3$  are the frequencies  
628 of genera with stratigraphic ranges of one, two, and three geologic time intervals,  
629 respectively (Foote and Raup 1996). The resulting estimate of 0.15-0.30 (Foote and Miller  
630 2007) thus represents the probability that one or more members of a given genus are  
631 sampled from a geological time interval, and Foote and Sepkoski (1999) used  
632 five-million-year time intervals in their analysis. As CladeAge calibration densities are  
633 calculated from instantaneous species-level sampling rates, we translated the genus-level  
634 sampling probability estimate of Foote and Sepkoski (1999) as follows. We downloaded the  
635 list of all valid scientific names of bony fishes from the Catalogue of Life database (Roskov

636 et al. 2015) and determined the frequency distribution of extant bony fish genus sizes from  
637 these names. We then used this distribution in combination with species-level sampling  
638 rates to simulate bony fish preservation over five million years and recorded the proportion  
639 of genera that were sampled during this interval. The species-level sampling rate was  
640 optimized until the resulting proportion of sampled genera was sufficiently close to the  
641 genus-level estimate of (Foote and Miller 2007). This optimization was performed  
642 separately for the lower and upper bound of estimate of (Foote and Miller 2007). We find  
643 that species-level instantaneous sampling rates of  $0.0066\text{-}0.01806 \text{ L}^{-1}\text{myr}^{-1}$  provide the  
644 best fit to five-million-year genus-level preservation probabilities of bony fishes (under the  
645 assumption of constant rates and a constant genus-size frequency distribution) and use this  
646 range of sampling rates for the calculation of CladeAge calibration densities. For  
647 comparison, and in order to provide species-level estimates for future users of CladeAge, we  
648 compiled a comprehensive list of published sampling rates in Supplementary Table S1,  
649 using the above translation where necessary.

### 650 *Divergence-Time Estimation of Teleost Fishes*

651 We analyzed the published fossil record for each of the 455 strongly supported  
652 teleost clades, and identified their first occurrences, the rock formation in which the earliest  
653 record was found, as well as the minimum and maximum age of this formation. Detailed  
654 information of the fossil record of each clade is given Supplementary Text S2. According to  
655 calibration scheme A, we used first occurrences to define CladeAge calibration densities  
656 distributions even if earlier records were known in sister clades, and we reused calibration  
657 densities for more inclusive clades if these (i) had no earlier fossil record on their own, but  
658 were (ii) either morphologically recognizable or characterized by a discrete geographical  
659 distribution so that fossil finds could in principle have been assigned to them directly  
660 rather than to parental clades only. For example, the Miocene *Nandopsis †woodringi*

661 represents the earliest record of the genus *Nandopsis*, to which it can be assigned based on  
662 the presence of lingual cusps on the oral teeth and four anal-fin spines, a character  
663 combination which within cichlids is unique to members of this genus (Chakrabarty 2007).  
664 However, *Nandopsis* †*woodringi* also represents the first occurrence of the clade  
665 “SCAC+NCAC”, combining the groups “SCAC” (Southern Central American Clade) and  
666 “NCAC” (Northern Central American Clade) of López-Fernández et al. (2010) with a total  
667 of 19 genera of Neotropical cichlids. This clade is strongly supported by molecular  
668 phylogenies (López-Fernández et al. 2010, this study), but is not characterized by known  
669 synapomorphies or a geographical distribution that separates it from its potential sister  
670 groups. Thus, if stem-group fossils were found of clade “SCAC+NCAC”, these would likely  
671 be misassigned to the next more inclusive clade that is morphologically recognizable, in  
672 this case the tribe Heroini. A lack of recognizable features for a clade thus effectively  
673 reduces its sampling rate to 0. In order to account for this reduction, CladeAge calibration  
674 densities were defined exclusively for clades that are morphologically (or in some cases  
675 geographically) recognizable. We identified fossil constraints for a total of 147 clades,  
676 including 18 clades within cichlids (see Supplementary Text S2 and Supplementary Figure  
677 S6).

678 In order to reduce model complexity and increase computational efficiency of  
679 Bayesian phylogenetic inference, eight markers with the greatest proportions of missing  
680 sequences were removed from the data set (Supplementary Table S4). In addition, a total  
681 of 80 codon positions with signatures of episodic selection were identified with the mixed  
682 effects model of evolution implemented in HyPhy (Murrell et al. 2012; Kosakovsky Pond  
683 et al. 2005) and removed from the alignment. We further collapsed each of the 362  
684 mutually exclusive clades to individual tips, and for each marker we chose sequences of  
685 clade members at random to represent the terminal clade. To account for sequence  
686 variation within a clade, we repeated random sequence sampling five times, producing five

687 replicate datasets that each included a total of 27 950 sites with 59.3% missing data. Each  
688 of the five replicate data sets was used for phylogenetic inference and time calibration with  
689 BEAST, on the basis of 147 CladeAge calibration densities. As for ML analyses, the data  
690 set was partitioned according to marker and codon position. Tree topology and branch  
691 lengths were linked among partitions, but parameters of the clock and sequence  
692 substitution models remained unlinked. We assumed an uncorrelated relaxed molecular  
693 clock (Drummond et al. 2006) and applied the reversible-jump based substitution model of  
694 Bouckaert et al. (2013). For each partition, a gamma distribution of among-site rate  
695 heterogeneity with four rate categories was assumed. We used the flexible birth-death  
696 skyline model (Stadler et al. 2012) with independent diversification rate parameters for the  
697 pre-Cretaceous Mesozoic ( $> 145.5$  Ma), the Early (145.5-99.6 Ma) and Late Cretaceous  
698 (99.6-66.0 Ma), as well as the Cenozoic ( $< 66.0$  Ma), and specified a sampling fraction  $\rho$  of  
699 0.0135 according to the ratio of tips included in the analysis to the total extant diversity of  
700 Clupeocephala. We left the tree topologically unconstrained except for nodes used for time  
701 calibration. Justifications for the assumed monophyly of each clade used for time  
702 calibration are given in Supplementary Text S2. For each data set replicate, 600 million  
703 MCMC states were sampled, and we repeated the analysis without data, sampling from the  
704 prior to ensure that conclusions were not pre-determined by the prior. Convergence was  
705 assessed by comparing MCMC traces among run replicates, and was verified by running  
706 additional analyses (650 million MCMC states) of the same model, but with substitution  
707 rate parameters fixed according to estimates obtained with jModelTest v.2.0 (Posada 2008).

### 708 *Resulting Timeline of Cichlid and Teleost Divergences*

709 Comparison of MCMC traces of the five run replicates suggested that all replicates  
710 had converged at the same posterior distribution, which was confirmed by the additional  
711 analyses with fixed substitution rate parameters. After discarding 60 million MCMC

712 generation of each replicate run as burn-in, we produced a joint posterior tree sample with  
713 1000 trees per replicate, and generated a Maximum Clade Credibility (MCC) tree from the  
714 combined distribution of 5000 posterior trees. The inferred timeline of cichlid and outgroup  
715 teleost divergences is summarized in Figure 4a and shown in more detail in Supplementary  
716 Figure S7.

717 The MCC tree topology was well supported, and corroborates the higher-level  
718 groupings found in recent large-scale Bayesian phylogenies of teleost fishes (Near et al.  
719 2013; Betancur-R et al. 2013), as well as previously identified relationships within cichlid  
720 fishes (e.g. Schwarzer et al. 2009; López-Fernández et al. 2010; Friedman et al. 2013; Meyer  
721 et al. 2015). With a single exception (Centropomidae), all unconstrained clades from our  
722 list of 455 clades were recovered as monophyletic. According to the MCC timeline, crown  
723 Clupeocephala originated around 207.8 Ma (95% HPD: 234.5-186.2 Ma), crown  
724 Acanthomorphata appeared 144.2 Ma (95% HPD: 158.4-130.6 Ma), and South American  
725 Cichlinae and African Pseudocrenilabrinae diverged about 81.6 Ma (95% HPD: 89.4-74.0  
726 Ma). In comparison, the age of crown Clupeocephala appears markedly older in the studies  
727 of Betancur-R et al. (2013) and Near et al. (2013), who estimated their origin at about  
728 251.1 (95% HPD: 276.1-226.1 Ma) and 273.7 Ma (95% HPD: 307.5-242.0 Ma), respectively.  
729 The divergence date of Acanthomorphata is more comparable between the three studies,  
730 and was estimated at 164.9 Ma (95% HPD 186.0-144.4 Ma) in Betancur-R et al. (2013),  
731 and around 142.5 Ma (95% HPD: 154.0-132.0 Ma) in Near et al. (2013), less than two  
732 million years younger than the estimate resulting from our time-calibrated phylogeny. For  
733 relatively younger divergences, however, our estimates appear older than those of  
734 Betancur-R et al. (2013), Near et al. (2013), and Friedman et al. (2013): The divergence of  
735 South American Cichlinae and African Pseudocrenilabrinae was estimated at 62.0 Ma (95%  
736 HPD: 70.4-53.9 Ma) in Betancur-R et al. (2013), at 46.4 Ma (95% HPD: 54.9-40.9 Ma) in  
737 Friedman et al. (2013), and as young as 26.0 Ma (95% HPD: 29.6-22.0 Ma) in Near et al.



738 (2013). Notably even the older limit of the 95% HPD of the latter estimate is predated by  
739 at least 5 well-characterized fossil species within crown Pseudocrenilabrinae (Murray  
740 2001b) and crown Cichlinae (Malabarba et al. 2010, 2006; Alano Perez et al. 2010;  
741 Malabarba and Malabarba 2008) and thus is in strong disagreement with the cichlid fossil  
742 record. Thus, the consistent application of CladeAge calibration densities to all clades with  
743 known fossil records appears to remove conflicts of comparatively younger node ages with  
744 the fossil record, while at the same time reducing the length of ghost lineages for relatively  
745 older clades. A more detailed comparison of clade age estimates between these studies is  
746 shown in Supplementary Figure S8.

747 While our age estimates for cichlid divergences are generally older than those  
748 obtained in Betancur-R et al. (2013) and Near et al. (2013), they are still markedly too  
749 young to support strictly Gondwanan vicariance between Indian, Malagassy, Neotropical,  
750 and African groups of cichlid fishes, as well as within other groups of freshwater fishes  
751 included in our phylogeny (Fig. 4). Notably, the divergence of Neotropical Cichlinae and  
752 African Pseudocrenilabrinae, estimated at 81.6 Ma, appears to have occurred about 20 myr  
753 after the final separation of the American and African landmasses at 104-100 Ma (Heine  
754 et al. 2013).

755 Comparison of these results with those obtained by MCMC sampling from the prior  
756 distribution shows that the divergence estimate for African and South American cichlids is  
757 driven by the molecular sequence data. The prior distribution is markedly older than the  
758 divergence date posterior for this split, with 65.6% of the prior samples being younger than  
759 100 Ma, whereas the same is true for 99.9% of the posterior distribution (Fig. 4d). As a  
760 consequence, the Bayes factor in favour of a divergence younger than 100 Ma is 752, which  
761 can be considered overwhelming evidence (Kass and Raftery 1995) supporting the  
762 trans-Atlantic dispersal scenario, as opposed to Gondwanan vicariance. Our results thus  
763 agree with those of Friedman et al. (2013), who found support for trans-Atlantic dispersal

764 based on three different approaches, including an analysis of the temporal distribution of  
765 cichlid-bearing fossil horizons and an analysis of the distribution of outgroup ages in  
766 addition to their time-calibrated phylogeny.

## 767 DISCUSSION

### 768 *Divergence-Time Estimation with CladeAge Calibration Densities*

769 Our analyses of datasets simulated under a wide range of conditions show that  
770 CladeAge calibration densities allow bias-free estimation of divergence times. The  
771 comparison of four different calibration schemes confirms that calibration scheme A (Fig.  
772 2a) performs better than other schemes and should thus be applied whenever CladeAge  
773 calibration densities are used for time calibration. This implies that for each clade, the  
774 oldest fossil record of this clade should be used as a time constraint, regardless of whether  
775 the fossil is also the earliest record of other (nested or parental) clades, and even if the  
776 fossil is younger than the oldest fossil record of the clade's sister group. With calibration  
777 scheme A, the CladeAge model produces very similar results to the FBD model with fixed  
778 sampling rates, and appears robust to model violation in the form of branch-rate  
779 autocorrelation or parameter misspecification by up to 50%, at least with larger numbers of  
780 fossil calibrations. We observe that when the sampling rate is not fixed in FBD analyses  
781 and only the oldest fossil of each clade is used for time calibration, the sampling rate is  
782 often substantially underestimated, leading to wide confidence intervals and overestimation  
783 of node ages. This suggests that when the sampling-rate parameter is not fixed in analyses  
784 with the FBD model, a rather complete representation of the fossil record should be  
785 included in the analysis instead of only the oldest fossil of each clade.

786 Other approaches have also been developed to utilize the complete known fossil  
787 record of clades for phylogenetic divergence-time estimation. Wilkinson et al. (2011) used  
788 approximate Bayesian computation to fit a model of speciation and preservation to the  
789 numbers of primate fossils known from different Paleocene epochs (Tavaré et al. 2002). The  
790 resulting posterior distributions for the ages of two clades were then used as calibration  
791 densities in a subsequent phylogenetic node dating analysis of primate sequence data. A  
792 more general model was developed by Nowak et al. (2013) to calculate the likelihood of a  
793 given “missing interval” (i.e. the duration between clade origin and its first preservation)  
794 based on diversification rates estimated from stratigraphic ranges of fossil taxa. The  
795 authors implemented this model in the software SNAPE, which further allows the fitting of  
796 parametric probability distributions to the calculated likelihoods, so that these probability  
797 distributions can then be used as calibration densities in Bayesian divergence-time  
798 estimation with BEAST or similar programs. While this method shares similarities with  
799 CladeAge, it differs from our approach in that it does not allow user-specified  
800 diversification rates, it does not account for uncertainty in rate estimates, and it requires  
801 information about the entire fossil record of a clade.

802 However, for the practical time calibration of higher-level phylogenies, the  
803 compilation of the entire fossil record of clades used for calibration may be far less feasible  
804 than the identification of their oldest reported fossils. A large amount of paleontological  
805 literature has been dedicated to determine oldest taxon appearances across the tree of life,  
806 and demonstrates the difficulties associated with the identification of these records as well  
807 as of their ages (Benton 1993; Benton and Donoghue 2007; Hedges and Kumar 2009;  
808 Ksepka et al. 2011; Benton et al. 2015). While the Paleobiological Database  
809 ([www.paleobiodb.org](http://www.paleobiodb.org)) provides information not only about the oldest fossils of clades, but  
810 about a much larger number of fossils for many clades, the taxonomic assignment and age  
811 ranges given for these fossils are usually far less well curated than those of first taxon

812 appearances that are dealt with in dedicated literature. In addition, neither the  
813 paleontological literature nor databases that use information from this literature are likely  
814 to provide an unbiased representation of the age distribution of fossils within a clade.  
815 Instead, new discoveries of fossils that extend the known age range of clades are almost  
816 guaranteed to be reported in the literature (and as a consequence also in databases),  
817 whereas younger findings may often not be considered worthy of publication. Thus,  
818 available information about the oldest records of clades is likely to be better curated and  
819 less biased than the collective data for all its fossils. Furthermore, since fossils are added as  
820 tips in analyses with the FBD model, the computational demand increases with the  
821 number of fossils, and may be prohibitive for higher-level phylogenies of clades with rich  
822 fossil records.

823         On the other hand, the specification of CladeAge calibration densities is  
824 computationally not more demanding than any other calibration density used in node  
825 dating, and is thus suitable for large-scale phylogenetic analyses. In contrast to the current  
826 software implementations of the FBD model (Gavryushkina et al. 2014; Zhang et al. 2016),  
827 the CladeAge method can also account for different sampling rates in different co-existing  
828 clades, as calibration densities are independently specified for each calibrated clade. This is  
829 likely to improve age estimates in higher-level phylogenies such as the vertebrate tree of  
830 life, where substantial differences in these rates have previously been demonstrated  
831 (summarized in Supplementary Table S1). Thus, a strategy for the Bayesian  
832 divergence-time estimation of large trees like the vertebrate tree of life could include the  
833 following steps. First, representative groups with suitable fossil records could be chosen  
834 from several of the higher taxa (i.e. mammals, birds, teleost fishes) included in the  
835 phylogeny, and could be used to estimate sampling rate parameters for these taxa. This  
836 could be done either using information from the fossil record alone (Silvestro et al. 2014;  
837 Starrfelt and Liow 2016), or in combination with molecular data, e.g. by means of separate

838 FBD analyses for each of the representative groups. Then, the resulting rate estimates  
839 could be used to calibrate the ages of clades within the higher taxa, under the assumption  
840 that the true rates of these clades are comparable to those estimated from representative  
841 groups. Finally, divergence-time estimation of the complete phylogeny could be carried out  
842 on the basis of CladeAge calibration densities.

### 843 *Trans-Atlantic Dispersal of Cichlid Fishes*

844 Using CladeAge calibration densities for 147 clades of teleost fishes, we found  
845 strong evidence for trans-oceanic dispersal, not only in cichlid fishes, but also in several  
846 other groups of freshwater fishes, including Cyprinodontoidei, Aplocheiloidei, and Siluroidei  
847 (Fig. 4). The calibration densities used in our analysis were based on estimates of sampling  
848 and diversification rates by Foote and Sepkoski (1999) and Santini et al. (2009), and our  
849 results could thus be biased if these estimates are inaccurate. We note, however, that the  
850 rate estimates used by us are low compared to those of other clades (see Supplementary  
851 Table S1) or those obtained by other authors (Rabosky et al. 2013). Thus, the used  
852 estimates for sampling and diversification rates are more likely underestimates than  
853 overestimates, which would lead to calibration densities that are wider than they should  
854 be, and therefore to overestimated ages of clades in our phylogeny.

855 For several further reasons, we would expect our age estimates to be rather over-  
856 than underestimated. First, our simulations have shown that age estimates obtained with  
857 CladeAge calibration densities (or the FBD model) can appear too old when the sampling  
858 rate is low and the assumed clock model is violated, e.g. by branch-rate autocorrelation  
859 (Supplementary Figure S3b). In practice, autocorrelation of branch-specific substitution  
860 rates can rarely be excluded, and may be present also in teleost fishes, as many factors  
861 influencing rate variation are heritable in vertebrates (Nabholz et al. 2008; Amster and  
862 Sella 2016). Second, while our molecular dataset is composed of both nuclear and

863 mitochondrial sequences, nuclear sequences were available to a greater degree for taxa  
864 outside of Cichlidae, and may be underrepresented for clades within this family. As the  
865 substitution rate of mitochondrial markers is usually higher than that of nuclear markers  
866 (Brown et al. 1979), overall genetic divergences between cichlids might appear higher than  
867 those of other groups that have a similar age but a lower proportion of missing data in  
868 nuclear markers. Third, by using concatenation of all sequence markers rather than the  
869 multispecies coalescent-model (which would have been computationally infeasible), we  
870 essentially ignored potential variation between gene trees due to incomplete lineage sorting,  
871 which has been shown to lead to inflated age estimates in several studies (McCormack  
872 et al. 2011; Colombo et al. 2015). Fourth, in contrast to the authors of a previous study on  
873 cichlid divergence times (Friedman et al. 2013), we assumed nested positions of the oldest  
874 Neotropical and African cichlid fossils within the subfamilies Cichlinae and  
875 Pseudocrenilabrinae, respectively, rather than positions in their stem groups. Specifically,  
876 we assumed a position within genus *Gymnogeophagus* for *Gymnogeophagus* †*eocenicus*, a  
877 position of †*Tremembichthys garciae* within Cichlasomatini, a position of †*Plesioheros*  
878 *chauliodus* within Heroini, a position of †*Proterocara argentina* within a clade formed by  
879 the extant genera *Teleocichla* and *Crenicichla*, and a position of †*Mahengechromis* spp.  
880 within the African tribe Hemichromini, which are all supported by morphological analyses  
881 (Murray 2000b, 2001b; Malabarba and Malabarba 2008; Smith et al. 2008; Alano Perez  
882 et al. 2010; Malabarba et al. 2010). If these nested positions should be unjustified (as  
883 suggested by Friedman et al. 2013), even younger ages of cichlid divergences would be  
884 expected.

885 On the other hand, we assumed the age of the earliest Neotropical cichlid fossils  
886 *Gymnogeophagus* †*eocenicus*, †*Plesioheros chauliodus*, and †*Proterocara argentina* to be  
887 slightly lower (45.4-39.9 Ma) than other authors (48.6 Ma; Alano Perez et al. 2010;  
888 Malabarba et al. 2010, 2014), as we consider the older age estimate for these fossils

889 unfounded and outdated (a detailed discussion of the age of these fossils is provided in  
890 Supplementary Text S2). However, even if these fossils were in fact 48.6 myr old, their age  
891 would not be in conflict with our time-calibrated phylogeny (Figure 4a, Supplementary  
892 Figure 7), according to which the genus *Gymnogeophagus* originated around 54.1 Ma (95%  
893 HPD: 63.0-45.3 Ma), the tribe Heroini appeared 67.5 Ma (95% HPD: 75.8-59.7 Ma), and  
894 the clade combining *Teleocichla* and *Crenicichla* originated 48.2 Ma (95% HPD: 56.3-41.3  
895 Ma). Therefore, we assume that using the age of 48.6 Ma for these fossils would have had  
896 negligible impact on the inferred timeline of cichlid diversification. Taken together, our  
897 analyses strongly support trans-Atlantic dispersal of cichlids, between 89.4 and 74.0 Ma, or  
898 earlier.

## 899 CONCLUSION

900 In this study, we have presented a new approach to Bayesian divergence-time  
901 estimation that is directly based on probabilities of fossil sampling, and thus overcomes  
902 previous shortcomings of node dating. We have demonstrated that our approach allows  
903 accurate and precise time calibration and represents a viable alternative to the FBD model  
904 when estimates for the rates of fossil sampling and diversification are available *a priori*.  
905 Our approach is particularly suitable for the time calibration of large-scale phylogenies,  
906 and we have outlined strategies how to use our method in order to account for variable  
907 rates of sampling and diversification in different clades. By applying our approach to a  
908 detailed phylogeny of teleost fishes, we have shown that freshwater fishes in several clades  
909 have diverged long after the separation of the continents on which they live, which implies  
910 that fishes from these clades have successfully traversed oceanic environments despite their  
911 adaptations to a freshwater lifestyle. These examples include the trans-Atlantic dispersal of  
912 cichlid fishes, which led to their colonization of South and Central American rivers and

913 lakes, and to the radiation of Neotropical cichlid fishes into over 600 extant species. We  
914 have implemented our approach in the CladeAge package for BEAST (Bouckaert et al.  
915 2014), which is freely available at [www.beast2.org](http://www.beast2.org).

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## 922 SUPPLEMENTARY MATERIAL

923 Supplementary material, including data files and online-only appendices, can be found in  
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925 Sequence data generated for this study has been deposited in GenBank (accession numbers  
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937 \*

## 938 References

939 Alano Perez, P., M. C. Malabarba, and C. Del Papa. 2010. A new genus and species of  
940 Heroini (Perciformes: Cichlidae) from the early Eocene of southern South America.  
941 *Neotrop. Ichthyol.* 8:631–642.

942 Alfaro, M. E., F. Santini, C. D. Brock, H. Alamillo, A. Dornburg, D. L. Rabosky,  
943 G. Carnevale, and L. J. Harmon. 2009. Nine exceptional radiations plus high turnover  
944 explain species diversity in jawed vertebrates. *Proc. Natl. Acad. Sci. USA*  
945 106:13410–13414.

946 Ali, J. R. and J. C. Aitchison. 2008. Gondwana to Asia: Plate tectonics, paleogeography  
947 and the biological connectivity of the Indian sub-continent from the Middle Jurassic  
948 through latest Eocene (166–35 Ma). *Earth-Sci. Rev.* 88:145–166.

949 Ali, J. R. and D. W. Krause. 2011. Late Cretaceous bioconnections between  
950 Indo-Madagascar and Antarctica: refutation of the Gunnerus Ridge causeway  
951 hypothesis. *J. Biogeogr.* 38:1855–1872.

952 Amster, G. and G. Sella. 2016. Life history effects on the molecular clock of autosomes and  
953 sex chromosomes. *Proc. Natl. Acad. Sci. USA* 113:1588–1593.

954 Arcila, D., R. A. Pyron, J. C. Tyler, G. Ortí, and R. Betancur-R. 2015. An evaluation of  
955 fossil tip-dating versus node-age calibrations in tetraodontiform fishes (Teleostei:  
956 Percomorphaceae). *Mol. Phylogenet. Evol.* 82:131–145.

- 957 Azuma, Y., Y. Kumazawa, M. Miya, K. Mabuchi, and M. Nishida. 2008. Mitogenomic  
958 evaluation of the historical biogeography of cichlids toward reliable dating of teleostean  
959 divergences. *BMC Evol. Biol.* 8:215.
- 960 Beaulieu, J. M. and M. J. Donoghue. 2013. Fruit evolution and diversification in  
961 campanulid angiosperms. *Evolution* 67:3132–3144.
- 962 Beck, R. M. D. and M. S. Y. Lee. 2014. Ancient dates or accelerated rates? Morphological  
963 clocks and the antiquity of placental mammals. *Proc. R. Soc. B* 281:20141278–20141278.
- 964 Benton, M. and P. Donoghue. 2007. Paleontological evidence to date the tree of life. *Mol.*  
965 *Biol. Evol.* 24:26–53.
- 966 Benton, M. J., ed. 1993. *The fossil record 2*. Chapman & Hall, London, UK.
- 967 Benton, M. J., M. J. Donoghue, R. J. Asher, M. Friedman, T. J. Near, and J. Vinther.  
968 2015. Constraints on the timescale of animal evolutionary history. *Palaeontol. Electron.*  
969 18.1.1FC:1–106.
- 970 Betancur-R., R. E. Broughton, E. O. Wiley, K. E. Carpenter, J. A. López, C. Li, N. I.  
971 Holcroft, D. Arcila, M. D. Sanciangco, J. C. Cureton, F. Zhang, T. Buser, M. A.  
972 Campbell, J. A. Ballesteros, A. Roa-Varon, S. C. Willis, W. C. Borden, T. Rowley, P. C.  
973 Reneau, D. J. Hough, G. Lu, T. Grande, G. Arratia, and G. Ortí. 2013. The Tree of Life  
974 and a new classification of bony fishes. *PLOS Currents: Tree of Life Pages* 1–45.
- 975 Bouckaert, R., M. V. Alvarado-Mora, and J. R. Rebelo Pinho. 2013. Evolutionary rates  
976 and HBV: issues of rate estimation with Bayesian molecular methods. *Antivir. Ther.*  
977 18:497–503.
- 978 Bouckaert, R., J. Heled, D. Kühnert, T. Vaughan, C.-H. Wu, D. Xie, M. A. Suchard,

979 A. Rambaut, and A. J. Drummond. 2014. BEAST 2: a software platform for Bayesian  
980 evolutionary analysis. *PLOS Comput. Biol.* 10:e1003537.

981 Bromham, L. and D. Penny. 2003. The modern molecular clock. *Nat. Rev. Genet.*  
982 4:216–224.

983 Brown, W. M., M. J. George, and A. C. Wilson. 1979. Rapid evolution of animal  
984 mitochondrial DNA. *Proc. Natl. Acad. Sci. USA* 76:1967–1971.

985 Chakrabarty, P. 2004. Cichlid biogeography: comment and review. *Fish Fish.* 5:97–119.

986 Chakrabarty, P. 2007. Taxonomic status of the hispaniolan Cichlidae. *Occas. Pap. Mus.*  
987 *Zool. Univ. Mich.* Pages 1–20.

988 Claramunt, S. and J. Cracraft. 2015. A new time tree reveals Earth history’s imprint on  
989 the evolution of modern birds. *Sci. Adv.* 1:e1501005.

990 Colombo, M., M. Damerau, R. Hanel, W. Salzburger, and M. Matschiner. 2015. Diversity  
991 and disparity through time in the adaptive radiation of Antarctic notothenioid fishes. *J.*  
992 *Evol. Biol.* 28:376–394.

993 Conkel, D. 1993. *Cichlids of North and Central America*. TFH Publications, Neptune City,  
994 New Jersey.

995 Cunningham, F., M. R. Amode, D. Barrell, K. Beal, K. Billis, S. Brent, D. Carvalho-Silva,  
996 P. Clapham, G. Coates, S. Fitzgerald, L. Gil, C. G. Giron, L. Gordon, T. Hourlier, S. E.  
997 Hunt, S. H. Janacek, N. Johnson, T. Juettemann, A. K. Kahari, S. Keenan, F. J. Martin,  
998 T. Maurel, W. McLaren, D. N. Murphy, R. Nag, B. Overduin, A. Parker, M. Patricio,  
999 E. Perry, M. Pignatelli, H. S. Riat, D. Sheppard, K. Taylor, A. Thormann, A. Vullo,  
1000 S. P. Wilder, A. Zadissa, B. L. Aken, E. Birney, J. Harrow, R. Kinsella, M. Muffato,

1001 M. Ruffier, S. M. J. Searle, G. Spudich, S. J. Trevanion, A. Yates, D. R. Zerbino, and  
1002 P. Flicek. 2015. Ensembl 2015. *Nucleic Acids Res.* 43:D662–D669.

1003 del Papa, C., A. Kirschbaum, J. Powell, A. Brod, F. Hongn, and M. Pimentel. 2010.  
1004 Sedimentological, geochemical and paleontological insights applied to continental  
1005 omission surfaces: A new approach for reconstructing an Eocene foreland basin in NW  
1006 Argentina. *J. South Am. Earth Sci.* 29:327–345.

1007 dos Reis, M., Y. Thawornwattana, K. Angelis, M. J. Telford, P. C. J. Donoghue, and  
1008 Z. Yang. 2015. Uncertainty in the timing of origin of animals and the limits of precision  
1009 in molecular timescales. *Curr. Biol.* 25:2939–2950.

1010 Drummond, A. J., S. Y. W. Ho, M. J. Philips, and A. Rambaut. 2006. Relaxed  
1011 phylogenetics and dating with confidence. *PLOS Biol.* 4:e88.

1012 Drummond, A. J., O. G. Pybus, A. Rambaut, R. Forsberg, and A. G. Rodrigo. 2003.  
1013 Measurably evolving populations. *Trends Ecol. Evol.* 18:481–488.

1014 Faria, N. R., A. Rambaut, M. A. Suchard, G. Baele, T. Bedford, M. J. Ward, A. J. Tatem,  
1015 J. D. Sousa, N. Arinaminpathy, J. Pépin, D. Posada, M. Peeters, O. G. Pybus, and  
1016 P. Lemey. 2014. The early spread and epidemic ignition of HIV-1 in human populations.  
1017 *Science* 346:56–61.

1018 Fernández, R., G. D. Edgecombe, and G. Giribet. 2016. Exploring phylogenetic  
1019 relationships within Myriapoda and the effects of matrix composition and occupancy on  
1020 phylogenomic reconstruction. *Syst. Biol.* Advance Access, doi: 10.1093/sysbio/syw041.

1021 Foote, M., J. P. Hunter, C. M. Janis, and J. J. Sepkoski, Jr. 1999. Evolutionary and  
1022 preservational constraints on origins of biologic groups: divergence times of eutherian  
1023 mammals. *Science* 283:1310–1314.

1024 Foote, M. and A. I. Miller. 2007. Principles of Paleontology. 3 ed. W. H. Freeman, New  
1025 York.

1026 Foote, M. and D. M. Raup. 1996. Fossil preservation and the stratigraphic ranges of taxa.  
1027 Paleobiology 22:121–140.

1028 Foote, M. and J. J. Sepkoski, Jr. 1999. Absolute measures of the completeness of the fossil  
1029 record. Nature 398:415–417.

1030 Forest, F. 2009. Calibrating the Tree of Life: fossils, molecules and evolutionary timescales.  
1031 Ann. Bot. 104:789–794.

1032 Friedman, M., B. P. Keck, A. Dornburg, R. I. Eytan, C. H. Martin, C. D. Hulsey, P. C.  
1033 Wainwright, and T. J. Near. 2013. Molecular and fossil evidence place the origin of  
1034 cichlid fishes long after Gondwanan rifting. Proc. R. Soc. B 280:20131733.

1035 Gavryushkina, A., T. A. Heath, D. T. Ksepka, T. Stadler, D. Welch, and A. J. Drummond.  
1036 2015. Bayesian total evidence dating reveals the recent crown radiation of penguins.  
1037 arXiv preprint Pages 1–65.

1038 Gavryushkina, A., D. Welch, T. Stadler, and A. J. Drummond. 2014. Bayesian inference of  
1039 sampled ancestor trees for epidemiology and fossil calibration. PLOS Comput. Biol.  
1040 10:e1003919.

1041 Genner, M. J., O. Seehausen, D. H. Lunt, D. A. Joyce, P. W. Shaw, G. R. Carvalho, and  
1042 G. F. Turner. 2007. Age of cichlids: new dates for ancient lake fish radiations. Mol. Biol.  
1043 Evol. 24:1269–1282.

1044 Gernhard, T. 2008. The conditioned reconstructed process. J. Theor. Biol. 253:769–778.

1045 Gire, S. K., A. Goba, K. G. Andersen, R. S. G. Sealfon, D. J. Park, L. Kanneh, S. Jalloh,  
1046 M. Momoh, M. Fullah, G. Dudas, S. Wohl, L. M. Moses, N. L. Yozwiak, S. Winnicki,

1047 C. B. Matranga, C. M. Malboeuf, J. Qu, A. D. Gladden, S. F. Schaffner, X. Yang, P.-P.  
1048 Jiang, M. Nekoui, A. Colubri, M. R. Coomber, M. Fonnies, A. Moigboi, M. Gbakie, F. K.  
1049 Kamara, V. Tucker, E. Konuwa, S. Saffa, J. Sellu, A. A. Jalloh, A. Kovoma, J. Koninga,  
1050 I. Mustapha, K. Kargbo, M. Foday, M. Yillah, F. Kanneh, W. Robert, J. L. B. Massally,  
1051 S. B. Chapman, J. Bochicchio, C. Murphy, C. Nusbaum, S. Young, B. W. Birren, D. S.  
1052 Grant, J. S. Scheiffelin, E. S. Lander, C. Happi, S. M. Gevao, A. Gnirke, A. Rambaut,  
1053 R. F. Garry, S. H. Khan, and P. C. Sabeti. 2014. Genomic surveillance elucidates Ebola  
1054 virus origin and transmission during the 2014 outbreak. *Science* 345:1369–1372.

1055 Greenfield, D. W. and J. E. Thomserson. 1997. *Fishes of the Continental Waters of Belize*.  
1056 University Press of Florida, Gainesville, Florida.

1057 Grimm, G. W., P. Kapli, B. Bomfleur, S. McLoughlin, and S. S. Renner. 2015. Using more  
1058 than the oldest fossils: dating Osmundaceae with three Bayesian clock approaches. *Syst.*  
1059 *Biol.* 64:396–405.

1060 Heath, T. A., J. P. Huelsenbeck, and T. Stadler. 2014. The fossilized birth-death process  
1061 for coherent calibration of divergence-time estimates. *Proc. Natl. Acad. Sci. USA*  
1062 111:E2957–E2966.

1063 Hedges, S. B. and S. Kumar, eds. 2009. *The Timetree of Life*. Oxford University Press,  
1064 Oxford, UK.

1065 Heine, C., J. Zoethout, and R. D. Müller. 2013. Kinematics of the South Atlantic rift. *Solid*  
1066 *Earth* 4:215–253.

1067 Heled, J. and A. J. Drummond. 2012. Calibrated tree priors for relaxed phylogenetics and  
1068 divergence time estimation. *Syst. Biol.* 61:138–149.

- 1069 Heled, J. and A. J. Drummond. 2015. Calibrated birth-death phylogenetic time-tree priors  
1070 for Bayesian inference. *Syst. Biol.* 64:369–383.
- 1071 Ho, S. Y. W. and M. J. Phillips. 2009. Accounting for calibration uncertainty in  
1072 phylogenetic estimation of evolutionary divergence times. *Syst. Biol.* 58:367–380.
- 1073 Höhna, S., T. Stadler, F. Ronquist, and T. Britton. 2011. Inferring speciation and  
1074 extinction rates under different sampling schemes. *Mol. Biol. Evol.* 28:2577–2589.
- 1075 Huelsenbeck, J. P. and F. Ronquist. 2001. MRBAYES: Bayesian inference of phylogenetic  
1076 trees. *Bioinformatics* 17:754–755.
- 1077 Jetz, W., G. H. Thomas, J. B. Joy, K. Hartmann, and A. Ø. Mooers. 2012. The global  
1078 diversity of birds in space and time. *Nature* 491:444–448.
- 1079 Kass, R. E. and A. E. Raftery. 1995. Bayes Factors. *J. Am. Stat. Assoc.* 90:773–795.
- 1080 Katoh, K. and D. M. Standley. 2013. MAFFT multiple sequence alignment software  
1081 version 7: improvements in performance and usability. *Mol. Biol. Evol.* 30:772–780.
- 1082 Klopstein, S., L. Vilhelmsen, and F. Ronquist. 2015. A nonstationary Markov model  
1083 detects directional evolution in hymenopteran morphology. *Syst. Biol.* 64:1089–1103.
- 1084 Kosakovsky Pond, S. L., S. D. W. Frost, and S. V. Muse. 2005. HyPhy: hypothesis testing  
1085 using phylogenies. *Bioinformatics* 21:676–679.
- 1086 Kosiol, C., I. Holmes, and N. Goldman. 2007. An empirical codon model for protein  
1087 sequence evolution. *Mol. Biol. Evol.* 24:1464–1479.
- 1088 Ksepka, D. T., M. J. Benton, M. T. Carrano, M. A. Gandolfo, J. J. Head, E. J. Hermsen,  
1089 W. G. Joyce, K. S. Lamm, J. S. L. Patané, M. J. Phillips, P. D. Polly, M. Van Tuinen,

- 1090 J. L. Ware, R. C. M. Warnock, and J. F. Parham. 2011. Synthesizing and databasing  
1091 fossil calibrations: divergence dating and beyond. *Biol. Lett.* 7:801–803.
- 1092 Lanfear, R., B. Calcott, S. Y. W. Ho, and S. Guindon. 2012. PartitionFinder: combined  
1093 selection of partitioning schemes and substitution models for phylogenetic analyses. *Mol.*  
1094 *Biol. Evol.* 29:1695–1701.
- 1095 Lepage, T., D. Bryant, H. Philippe, and N. Lartillot. 2007. A general comparison of relaxed  
1096 molecular clock models. *Mol. Biol. Evol.* 24:2669–2680.
- 1097 Lepage, T., S. Lawi, P. Tupper, and D. Bryant. 2006. Continuous and tractable models for  
1098 the variation of evolutionary rates. *Math. Biosci.* 199:216–233.
- 1099 Lewis, P. O. 2001. A likelihood approach to estimating phylogeny from discrete  
1100 morphological character data. *Syst. Biol.* 50:913–925.
- 1101 López-Fernández, H., K. O. Winemiller, and R. L. Honeycutt. 2010. Multilocus phylogeny  
1102 and rapid radiations in Neotropical cichlid fishes (Perciformes: Cichlidae: Cichlinae).  
1103 *Mol. Phylogenet. Evol.* 55:1070–1086.
- 1104 Malabarba, M. C. and L. R. Malabarba. 2008. A new cichlid *Tremembichthys garcia*  
1105 (Actinopterygii, Perciformes) from the Eocene-Oligocene of Eastern Brazil. *Rev. Bras.*  
1106 *Paleontol.* 11:59–68.
- 1107 Malabarba, M. C., L. R. Malabarba, and C. Del Papa. 2010. *Gymnogeophagus eocenicus*, n.  
1108 sp (Perciformes: Cichlidae), an Eocene cichlid from the Lumbrera formation in  
1109 Argentina. *J. Vert. Paleontol.* 30:341–350.
- 1110 Malabarba, M. C., L. R. Malabarba, and H. López-Fernández. 2014. On the Eocene  
1111 cichlids from the Lumbrera Formation: additions and implications for the Neotropical  
1112 ichthyofauna. *J. Vert. Paleontol.* 34:49–58.



- 1113 Malabarba, M. C., O. Zuleta, and C. Del Papa. 2006. *Proterocara argentina*, a new fossil  
1114 cichlid from the Lumbreira Formation, Eocene of Argentina. *J. Vert. Paleontol.*  
1115 26:267–275.
- 1116 Marshall, C. R. 2008. A simple method for bracketing absolute divergence times on  
1117 molecular phylogenies using multiple fossil calibration points. *Am. Nat.* 171:726–742.
- 1118 Martin, A. P. and S. R. Palumbi. 1993. Body size, metabolic rate, generation time, and the  
1119 molecular clock. *Proc. Natl. Acad. Sci. USA* 90:4087–4091.
- 1120 McCormack, J. E., J. Heled, K. S. Delaney, A. T. Peterson, and L. L. Knowles. 2011.  
1121 Calibrating divergence times on species trees versus gene trees: implications for  
1122 speciation history of *Aphelocoma* jays. *Evolution* 65:184–202.
- 1123 McMahan, C. D., P. Chakrabarty, J. S. Sparks, W. L. Smith, and M. P. Davis. 2013.  
1124 Temporal patterns of diversification across global cichlid biodiversity (Acanthomorpha:  
1125 Cichlidae). *PLOS ONE* 8:e71162.
- 1126 Meyer, B. S., M. Matschiner, and W. Salzburger. 2015. A tribal level phylogeny of Lake  
1127 Tanganyika cichlid fishes based on a genomic multi-marker approach. *Mol. Phylogenet.*  
1128 *Evol.* 83:56–71.
- 1129 Murray, A. M. 2000a. Eocene cichlid fishes from Tanzania, East Africa. *J. Vert. Paleontol.*  
1130 20:651–664.
- 1131 Murray, A. M. 2000b. The Eocene cichlids (Perciformes: Labroidei) of Mahenge, Tanzania.  
1132 Ph.D. thesis McGill University Montreal.
- 1133 Murray, A. M. 2001a. The fossil record and biogeography of the Cichlidae (Actinopterygii:  
1134 Labroidei). *Biol. J. Linn. Soc.* 74:517–532.

- 1135 Murray, A. M. 2001b. The oldest fossil cichlids (Teleostei: Perciformes): indication of a 45  
1136 million-year-old species flock. *Proc. R. Soc. B* 268:679–684.
- 1137 Murrell, B., J. O. Wertheim, S. Moola, T. Weighill, K. Scheffler, and S. L.  
1138 Kosakovsky Pond. 2012. Detecting individual sites subject to episodic diversifying  
1139 selection. *PLOS Genet.* 8:e1002764.
- 1140 Myers, G. S. 1949. Salt-tolerance of fresh-water fish groups in relation to zoogeographical  
1141 problems. *Bijdragen tot de Dierkunde* 28:315–322.
- 1142 Nabholz, B., S. Glemin, and N. Galtier. 2008. Strong variations of mitochondrial mutation  
1143 rate across mammals - the longevity hypothesis. *Mol. Biol. Evol.* 25:120–130.
- 1144 Near, T. J., A. Dornburg, R. I. Eytan, B. P. Keck, W. L. Smith, K. L. Kuhn, J. A. Moore,  
1145 S. A. Price, F. T. Burbrink, M. Friedman, and P. C. Wainwright. 2013. Phylogeny and  
1146 tempo of diversification in the superradiation of spiny-rayed fishes. *Proc. Natl. Acad. Sci.*  
1147 *USA* 110:12738–12743.
- 1148 Nowak, M. D., A. B. Smith, C. Simpson, and D. J. Zwickl. 2013. A simple method for  
1149 estimating informative node age priors for the fossil calibration of molecular divergence  
1150 time analyses. *PLOS ONE* 8:e66245.
- 1151 O'Reilly, J. E., M. dos Reis, and P. C. J. Donoghue. 2015. Dating tips for divergence-time  
1152 estimation. *Trends Genet.* 31:637–650.
- 1153 Orlando, L., A. Ginolhac, G. Zhang, D. Froese, A. Albrechtsen, M. Stiller, M. Schubert,  
1154 E. Cappellini, B. Petersen, I. Moltke, P. L. F. Johnson, M. Fumagalli, J. T. Vilstrup,  
1155 M. Raghavan, T. Korneliussen, A.-S. Malaspinas, J. Vogt, D. Szklarczyk, C. D. Kelstrup,  
1156 J. Vinther, A. Dolocan, J. Stenderup, A. M. V. Velazquez, J. Cahill, M. Rasmussen,  
1157 X. Wang, J. Min, G. D. Zazula, A. Seguin-Orlando, C. Mortensen, K. Magnussen, J. F.

1158 Thompson, J. Weinstock, K. Gregersen, K. H. Røed, V. Eisenmann, C. J. Rubin, D. C.  
1159 Miller, D. F. Antczak, M. F. Bertelsen, S. Brunak, K. A. S. Al-Rasheid, O. Ryder,  
1160 L. Andersson, J. Mundy, A. Krogh, M. T. P. Gilbert, K. Kjær, T. Sicheritz-Ponten, L. J.  
1161 Jensen, J. V. Olsen, M. Hofreiter, R. Nielsen, B. Shapiro, J. Wang, and E. Willerslev.  
1162 2014. Recalibrating *Equus* evolution using the genome sequence of an early Middle  
1163 Pleistocene horse. *Nature* 498:74–78.

1164 Pfeiffer, W. and A. Stamatakis. 2010. Hybrid MPI/Pthreads parallelization of the RAxML  
1165 phylogenetics code. Ninth IEEE International Workshop on High Performance  
1166 Computational Biology (HiCOMB 2010), Atlanta Pages 1–8.

1167 Posada, D. 2008. jModelTest: phylogenetic model averaging. *Mol. Biol. Evol.* 25:1253–1256.  
1168

1169 Prum, R. O., J. S. Berv, A. Dornburg, D. J. Field, J. P. Townsend, E. M. Lemmon, and  
1170 A. R. Lemmon. 2015. A comprehensive phylogeny of birds (*Aves*) using targeted  
1171 next-generation DNA sequencing. *Nature* 526:569–573.

1172 Pyron, R. A. 2011. Divergence time estimation using fossils as terminal taxa and the  
1173 origins of Lissamphibia. *Syst. Biol.* 60:466–481.

1174 Rabosky, D. L. 2014. Automatic detection of key innovations, rate shifts, and  
1175 diversity-dependence on phylogenetic trees. *PLOS ONE* 9:e89543.

1176 Rabosky, D. L., F. Santini, J. Eastman, S. A. Smith, B. Sidlauskas, J. Chang, and M. E.  
1177 Alfaro. 2013. Rates of speciation and morphological evolution are correlated across the  
1178 largest vertebrate radiation. *Nat. Commun.* 4:1958.

1179 Ratnasingham, S. and P. D. N. Hebert. 2007. BOLD: The Barcode of Life Data System  
1180 ([www.barcodinglife.org](http://www.barcodinglife.org)). *Mol. Ecol. Notes* 7:355–364.

- 1181 Ronquist, F. and J. P. Huelsenbeck. 2003. MrBayes 3: Bayesian phylogenetic inference  
1182 under mixed models. *Bioinformatics* 19:1572–1574.
- 1183 Ronquist, F., S. Klopfstein, L. Vilhelmsen, S. Schulmeister, D. L. Murray, and A. P.  
1184 Rasnitsyn. 2012. A total-evidence approach to dating with fossils, applied to the early  
1185 radiation of the Hymenoptera. *Syst. Biol.* 61:973–999.
- 1186 Roskov, Y., L. Abucay, T. Orell, D. Nicolson, T. Kunze, C. Flann, N. Bailly, P. Kirk,  
1187 T. Bourgoin, R. E. DeWalt, W. Decock, and A. De Wever. 2015. Species 2000 & ITIS  
1188 Catalogue of Life. Digital resource at [www.catalogueoflife.org/col](http://www.catalogueoflife.org/col). Species 2000:  
1189 Naturalis, Leiden, the Netherlands.
- 1190 Salzburger, W., B. Van Bocxlaer, and A. S. Cohen. 2014. Ecology and evolution of the  
1191 African Great Lakes and their faunas. *Annu. Rev. Ecol. Evol. Syst.* 45:519–545.
- 1192 Santini, F., L. J. Harmon, G. Carnevale, and M. E. Alfaro. 2009. Did genome duplication  
1193 drive the origin of teleosts? A comparative study of diversification in ray-finned fishes.  
1194 *BMC Evol. Biol.* 9:194.
- 1195 Sarich, V. M. and A. C. Wilson. 1967. Immunological time scale for hominid evolution.  
1196 *Science* 158:1200–1203.
- 1197 Schwarzer, J., B. Misof, D. Tautz, and U. K. Schlieven. 2009. The root of the East African  
1198 cichlid radiations. *BMC Evol. Biol.* 9:186.
- 1199 Silvestro, D., N. Salamin, and J. Schnitzler. 2014. PyRate: a new program to estimate  
1200 speciation and extinction rates from incomplete fossil data. *Method Ecol. Evol.*  
1201 5:1126–1131.
- 1202 Smith, G. J. D., D. Vijaykrishna, J. Bahl, S. J. Lycett, M. Worobey, O. G. Pybus, S. K.  
1203 Ma, C. L. Cheung, J. Raghwani, S. Bhatt, J. S. M. Peiris, Y. Guan, and A. Rambaut.

- 1204 2009. Origins and evolutionary genomics of the 2009 swine-origin H1N1 influenza A  
1205 epidemic. *Nature* 459:1122–1125.
- 1206 Smith, W. L., P. Chakrabarty, and J. S. Sparks. 2008. Phylogeny, taxonomy, and evolution  
1207 of Neotropical cichlids (Teleostei: Cichlidae: Cichlinae). *Cladistics* 24:625–641.
- 1208 Sparks, J. S. and W. L. Smith. 2004. Phylogeny and biogeography of cichlid fishes  
1209 (Teleostei: Perciformes: Cichlidae). *Cladistics* 20:501–517.
- 1210 Sparks, J. S. and W. L. Smith. 2005. Freshwater fishes, dispersal ability, and nonevidence:  
1211 “Gondwana life rafts” to the rescue. *Syst. Biol.* 54:158–165.
- 1212 Stadler, T. 2010. Sampling-through-time in birth–death trees. *J. Theor. Biol.* 267:396–404.
- 1213 Stadler, T. 2011. Mammalian phylogeny reveals recent diversification rate shifts. *Proc.*  
1214 *Natl. Acad. Sci. USA* 108:6187–6192.
- 1215 Stadler, T. and F. Bokma. 2013. Estimating speciation and extinction rates for phylogenies  
1216 of higher taxa. *Syst. Biol.* 62:220–230.
- 1217 Stadler, T., D. Kühnert, S. Bonhoeffer, and A. J. Drummond. 2012. Birth–death skyline  
1218 plot reveals temporal changes of epidemic spread in HIV and hepatitis C virus (HCV).  
1219 *Proc. Natl. Acad. Sci. USA* 110:228–233.
- 1220 Stamatakis, A. 2006. RAxML-VI-HPC: maximum likelihood-based phylogenetic analyses  
1221 with thousands of taxa and mixed models. *Bioinformatics* 22:2688–2690.
- 1222 Stamatakis, A., P. Hoover, and J. Rougemont. 2008. A rapid bootstrap algorithm for the  
1223 RAxML web servers. *Syst. Biol.* 57:758–771.
- 1224 Starrfelt, J. and L. H. Liow. 2016. How many dinosaur species were there? Fossil bias and

1225 true richness estimated using a Poisson sampling model. *Phil. Trans. R. Soc. B*  
1226 371:20150219.

1227 Stickney, R. R. 1986. Tilapia tolerance of saline waters: a review. *Prog. Fish Cult.* 48.

1228 Tavaré, S., C. R. Marshall, O. Will, C. Soligo, and R. D. Martin. 2002. Using the fossil  
1229 record to estimate the age of the last common ancestor of extant primates. *Nature*  
1230 416:726–729.

1231 Uchida, K., T. Kaneko, H. Miyazaki, S. Hasegawa, and T. Hirano. 2000. Excellent salinity  
1232 tolerance of Mozambique tilapia (*Oreochromis mossambicus*): elevated chloride cell  
1233 activity in the branchial and opercular epithelia of the fish adapted to concentrated  
1234 seawater. *Zool. Sci.* 17:149–160.

1235 Vences, M., J. Freyhof, R. Sonnenberg, J. Kosuch, and M. Veith. 2001. Reconciling fossils  
1236 and molecules: Cenozoic divergence of cichlid fishes and the biogeography of  
1237 Madagascar. *J. Biogeogr.* 28:1091–1099.

1238 Vucetich, M. G., M. A. Reguero, M. Bond, A. M. Candela, A. A. Carlini, C. M.  
1239 Deschamps, J. N. Gelfo, F. J. Goin, G. M. López, E. Ortiz Jaureguizar, R. Pascual, G. J.  
1240 Scillato-Yané, and E. C. Vieytes. 2007. Mamíferos continentales del Paleógeno argentino:  
1241 las investigaciones de los últimos cincuenta años. *Ameghiniana Publicación Especial*  
1242 11:239–255.

1243 Welker, F., M. J. Collins, J. A. Thomas, M. Wadsley, S. Brace, E. Cappellini, S. T. Turvey,  
1244 M. Reguero, J. N. Gelfo, A. Kramarz, J. Burger, J. Thomas-Oates, D. A. Ashford, P. D.  
1245 Ashton, K. Rowsell, D. M. Porter, B. Kessler, R. Fischer, C. Baessmann, S. Kaspar, J. V.  
1246 Olsen, P. Kiley, J. A. Elliott, C. D. Kelstrup, V. Mullin, M. Hofreiter, E. Willerslev, J.-J.  
1247 Hublin, L. Orlando, I. Barnes, and R. D. E. MacPhee. 2015. Ancient proteins resolve the  
1248 evolutionary history of Darwin’s South American ungulates. *Nature* 522:81–84.

- 1249 Wilkinson, R. D., M. E. Steiper, C. Soligo, R. D. Martin, Z. Yang, and S. Tavaré. 2011.  
1250 Dating primate divergences through an integrated analysis of palaeontological and  
1251 molecular data. *Syst. Biol.* 60:16–31.
- 1252 Wilson, A. C., S. S. Carlson, and T. J. White. 1977. Biochemical evolution. *Annu. Rev.*  
1253 *Biochem.* 46:573–639.
- 1254 Wright, A. M., G. T. Lloyd, and D. M. Hillis. 2016. Modeling character change  
1255 heterogeneity in phylogenetic analyses of morphology through the use of priors. *Syst.*  
1256 *Biol.* 65:602–611.
- 1257 Yang, Z. and B. Rannala. 2006. Bayesian estimation of species divergence times under a  
1258 molecular clock using multiple fossil calibrations with soft bounds. *Mol. Biol. Evol.*  
1259 23:212–226.
- 1260 Zhang, C., T. Stadler, S. Klopstein, T. A. Heath, and F. Ronquist. 2016. Total-evidence  
1261 dating under the fossilized birth–death process. *Syst. Biol.* 65:228–249.
- 1262 Zuckerkandl, E. and L. Pauling. 1962. Molecular disease, evolution, and genic  
1263 heterogeneity. Pages 189–225 *in* *Horizons in Biochemistry* (M. Kasha and B. Pullman,  
1264 eds.). Academic Press, New York.

Figure 1: Exemplary CladeAge calibration densities.

Probability densities for the age of a clade for which the earliest fossil is known to be exactly 10 myr old (a), or assumed to be between 10 and 30 myr old, with a uniform fossil age probability within this range (b). The gray area in b) indicates the fossil age uncertainty. Speciation rate and extinction rates are assumed to be  $\lambda = 0.08$  and  $\mu = 0.04$ , and sampling rates  $\psi$  are as indicated.

Figure 2: Four alternative calibration schemes for CladeAge calibration densities.

a) Assume that fossils  $F_1$  and  $F_2$ , represented by white circles, are the oldest fossil records of clades A and B, respectively, (here, they are part of the stem groups of these clades), and that no fossils are known outside of clades A and B. Fossil  $F_1$  can then be used to constrain the age of origin of clade A, marked with a black dot and the label  $O_1$ , while  $F_2$  can be used to constrain the age of origin of clade B, marked with the label  $O_2$ . As clades A and B are sister lineages,  $O_1$  and  $O_2$  are identical in age, which means that calibration densities for  $O_1$  and  $O_2$  would directly interact with each other. Further assuming that  $F_2$  is older than  $F_1$ ,  $F_2$  represents the first fossil record not only of clade B, but also of the more inclusive clades C and D. It could thus also be used to constrain the age of origin of these two clades, indicated by the labels  $O_3$  and  $O_4$ , respectively. In scheme A, each fossil is used to constrain the age of origin of all clades for which this fossils represents the earliest record. In schemes B and C, each fossil is used to constrain only the age of origin of the most inclusive clade (scheme B), or only the age of origin of the least inclusive clade (scheme C), for which it represents the first occurrence (as scheme B would otherwise allow  $O_2$  and  $O_3$  to be younger than  $F_2$ , we combine CladeAge calibration densities for this scheme always with hard lower bounds defined by the fossil age; see main text). Scheme D is similar to scheme C except that only the older one of two fossils in two sister clades is used as an age constraint. b) Comparison of waiting times between clade origin and first fossil occurrence. Waiting times between clade origin and first fossil occurrence were recorded from 10 000 simulated phylogenies with three different sampling rates ( $\psi = 0.1$ ,  $\psi = 0.03$ ,  $\psi = 0.01$ ), using schemes A-D, and a clade age threshold of 0.9. The frequency distributions of binned waiting times are shown in gray, and CladeAge probability density distributions for the same settings are indicated with dashed black lines. The total number of waiting times sampled is given in each plot.



Figure 3: Estimates of node ages in simulated phylogenies, obtained with four CladeAge calibration schemes and the FBD model.

Results are based on 50 simulated phylogenetic trees and sequence data, and fossil records simulated with three different sampling rates for each phylogeny. a) The mean number of fossil constraints used with each scheme, sorted into bins of 20 time units according to fossil age. For schemes B, C, and the FBD model, this number is identical to the number of fossils. In scheme A, some fossils are used for multiple constraints, and in scheme D, not all fossils are used (see Fig. 2). b) Estimated node ages with MCMC sampling from the prior alone, when the fossil record was simulated with the intermediate sampling rate  $\psi = 0.03$ . Node age comparisons based on other sampling rates ( $\psi = 0.1$  or  $\psi = 0.01$ ) are shown in Supplementary Figure S3. c) As b, but using MCMC sampling from the posterior, with sequence data generated with uncorrelated branch rates. Results for data sets with autocorrelated branch rates are shown in Figure S3. d) Mean width of 95% HPD intervals, when using MCMC sampling from the posterior with datasets generated with uncorrelated branch rates. Results are given in bins of 20 time units according to the true node age. e) Percentage of age estimates for which the 95% HPD interval includes the true node age, when sampling from the posterior and using datasets generated with uncorrelated branch rates. As in d), results are presented in bins of 20 time units according to the true node age. See Supplementary Tables S10-S12 for summary statistics for the full set of analyses.

Figure 4: Time-calibrated phylogenetic tree of teleost fishes and plate tectonic reconstructions.

a) Maximum Clade Credibility phylogeny of cichlid and outgroup teleost fishes, time-calibrated with 147 fossil constraints. Dashed lines mark continental break-up events of Gondwanan landmasses. Colors of terminal branches indicate the center of diversity for clades occurring exclusively in freshwater or brackish water habitats, using the same color code as in b). Groups with marine representatives are shown in light gray. Colors of internal branches indicate past distributions according to the most parsimonious scenario of dispersal and freshwater colonization, taking into account past geographic distances between landmasses. Black branches indicate equal parsimony of multiple scenarios. Six dispersal events with particularly strong evidence for trans-oceanic dispersal are highlighted: 1) Since the two oldest cichlid subfamilies, Etroplinae and Ptychochrominae, occur on Madagascar (and Ptychochrominae being endemic to Madagascar), this landmass represents the most likely origin of family Cichlidae. According to our timeline of teleost divergences, dispersal of the clade combining the younger two subfamilies Pseudocrenilabrinae and Cichlinae from Madagascar to either Africa or South America occurred after 85.7 Ma (95% HPD: 93.8-77.8 Ma), substantially later than the latest possible separation of Madagascar from both landmasses around 120 Ma (Ali and Aitchison 2008; Ali and Krause 2011). Since Madagascar was geographically closer to Africa than to South America at 85.7 Ma, we assume that cichlids dispersed to Africa before reaching South America. 2) The divergence event of African Pseudocrenilabrinae and South American Cichlinae is estimated at 81.8 Ma (95% HPD: 89.4-74.0 Ma), long after the final separation of the two continents at 104-100 Ma (Heine et al. 2013). 3) Within the cichlid subfamily Etroplinae, the Indian genus *Etroplus* and the Malagassy genus *Paretroplus* diverged about 69.5 Ma (95% HPD: 85.9-53.1 Ma), probably after the break-up of India and Madagascar between 90-85 Ma (Ali and Aitchison 2008). 4) The predominantly American Cyprinodontoidei include multiple Old World lineages, such as the clade combining the Mediterranean *Aphanius* and Valenciidae, which diverged from South American relatives about 50.6 Ma (95% HPD: 61.4-39.3 Ma). 5) With an estimated crown age 80.8 Ma (95% HPD: 92.5-69.7 Ma), the cyprinodontiform suborder Aplocheiloidei includes American, African, Malagassy, and Indian lineages of strict freshwater fishes. The aplocheilid sister genera *Pachypanchax* and *Aplocheilus* occur in Madagascar and Asia, respectively, and diverged about 42.8 Ma (95% HPD: 60.4-23.8 Ma). 6) The Mexican *Lacantunia enigmatica* appears deeply nested within African freshwater Siluroidei, but separated about 49.6 Ma (95% HPD: 57.9-45.2 Ma). b) Plate tectonic reconstructions of the break-up of Gondwana between 200 Ma and the present.

Figure 4: c) Stages in the separation of South America and Africa between 118 and 100 Ma. According to the plate kinematic model of Heine et al. (2013), final breakup in the South Atlantic Rift System (SARS) occurred between 113-112 Ma in the outer Santos Basin. African and South American lithospheres completely separated at 104 Ma, whereby the last continental connection remained along the Côte d'Ivoire/Ghanaian Ridge in the Equatorial Rift System (EqRS). Colored outlines represent Africa and South America with present coastlines. Dark gray shapes indicate the restored continental margin (see Heine et al. 2013). Modified from Heine et al. (2013). d) Prior and posterior distributions for the divergence date of African and South American cichlid fishes (marked with "2" in a). 99.9% of the posterior probability mass supports a divergence event younger than 100 Ma, and thus trans-Atlantic dispersal instead of Gondwanan vicariance. In contrast, this scenario is supported by only 65.6% of the prior probability.