

Pulled Diversification Rates, Lineages-Through-Time Plots and Modern Macroevolutionary Modelling

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ABSTRACT

1 Estimating time-dependent rates of speciation and extinction from dated phylogenetic
2 trees of extant species (timetrees), and determining how and why they vary, is key to
3 understanding how ecological and evolutionary processes shape biodiversity. Due to an
4 increasing availability of phylogenetic trees, a growing number of process-based methods
5 relying on the birth-death model have been developed in the last decade to address a
6 variety of questions in macroevolution. However, this methodological progress has regularly
7 been criticised such that one may wonder how reliable the estimations of speciation and
8 extinction rates are. In particular, using lineages-through-time (LTT) plots, a recent study
9 (Louca and Pennell, 2020) has shown that there are an infinite number of equally likely
10 diversification scenarios that can generate any timetree. This has lead to questioning
11 whether or not diversification rates should be estimated at all. Here we summarize, clarify,
12 and highlight technical considerations on recent findings regarding the capacity of models
13 to disentangle diversification histories. Using simulations we demonstrate the
14 characteristics of newly-proposed "pulled rates" and their utility. We recognize that the
15 recent findings are a step forward in understanding the behavior of macroevolutionary
16 modelling, but they in no way suggest we should abandon diversification modelling
17 altogether. On the contrary, the study of macroevolution using phylogenetic trees has
18 never been more exciting and promising than today. We still face important limitations in
19 regard to data availability and methodological shortcomings, but by acknowledging them
20 we can better target our joint efforts as a scientific community.

21 *Key words:* Birth-death models, extinction, speciation, phylogenetics

23

INTRODUCTION

24 A major goal in evolutionary biology is to understand the large-scale processes that
25 have shaped biodiversity patterns through time. One important way to investigate this is
26 by modelling species diversification using speciation and extinction, which can vary over
27 time and among groups. It is commonplace to find areas, or clades, in phylogenetic trees
28 that accumulate lineages faster than others. Diversification models often aim to explain
29 this variation in diversification patterns by associating bursts of speciation or extinction
30 with factors such as time (Höhna et al., 2016b), lineages (Rabosky, 2014), character traits
31 (Maddison et al., 2007), or the environment (Condamine et al., 2013).

32 The growing number of large phylogenetic trees that capture a significant
33 proportion of living species provide increasing power and resolution for such studies (Jetz
34 et al., 2012; Smith and Brown, 2018; Upham et al., 2019). Furthermore, the availability of
35 a wide variety of methods and software (e.g. BAMM (Rabosky, 2014), state-speciation and
36 extinction (SSE) models (Maddison et al., 2007), RPANDA (Morlon et al., 2016),
37 MEDUSA (Alfaro et al., 2009)) have made diversification studies increasingly popular in
38 the last decade. Approaches that can link diversification to a particular process or trait are
39 among the most appealing to researchers in the field because they enable us to test
40 long-standing hypotheses in evolutionary biology and ecology. Examples include those
41 related to the evolution of key innovations (Silvestro et al., 2014), the colonisation of new
42 areas (McGuire et al., 2014), the effect of elevation (Lagomarsino et al., 2016; Quintero
43 and Jetz, 2018) and the latitudinal diversity gradient (Rolland et al., 2014;
44 Pulido-Santacruz and Weir, 2016; Rabosky et al., 2018; Igea and Tanentzap, 2020).

45 A recent study ((Louca and Pennell, 2020) abbreviated to LP) demonstrates how
46 one approach, based on lineages-through-time (LTT) plots, cannot reliably estimate rates
47 of speciation and extinction over time using extant timetrees. LP show how results of this
48 approach can be misleading and provide potential solutions to the issues raised by
49 proposing new summary statistics. This publication has already provoked a response from

50 the community (e.g. Morlon et al. (2020)) and stimulated considerable discussion, with
51 some going so far as to suggest that speciation and extinction rates cannot be estimated
52 using phylogenetic trees (Pagel, 2020). As a result, this study has called into question the
53 meaning of diversification rate estimates generated from any analytical framework. Here,
54 we aim to outline the major concepts discussed in LP in an accessible way, targeting a
55 broad audience. We then put the results and conclusions of LP into historical context and
56 explore how the implications of this study apply to macroevolutionary modelling today.

57 MODELLING DIVERSIFICATION RATES

58 A typical workflow for diversification rate modelling using molecular phylogenetic
59 trees is as follows. DNA sequence data are obtained for species in a study group, which are
60 then used to estimate species relationships in the form of a phylogenetic tree. Typically,
61 this phylogenetic tree contains only extant species, and it is time-calibrated using ages
62 derived from different sources including fossils (Sauquet, 2013; Ho and Phillips, 2009). The
63 output of this process is referred to as an extant timetree. Once a tree has been generated,
64 a birth-death model is fitted to explain patterns of diversification in the tree. Note,
65 however, that fossils are usually used for node calibration and tree shape estimation but
66 are rarely incorporated in subsequent estimation of diversification rates, although recent
67 methodological progresses now allow incorporating fossils as tips in the phylogenies
68 (Ronquist et al., 2012; Heath et al., 2014) and birth-death models allow estimating rates of
69 diversification Mitchell et al. (2019).

70 The simplest birth-death models assume that each branch of a phylogenetic tree
71 shares the same rate of "birth" (speciation) events, as well as "death" (extinction) events
72 (Nee et al., 1994; Nee, 2006; Ricklefs, 2007; Morlon et al., 2011). There are two principal
73 parameters in the birth-death model, the speciation rate (λ) - the rate at which lineages
74 arise, and the extinction rate (μ) - the rate at which lineages disappear. Under this simple
75 framework λ and μ are constant over time (time-independent) and the same across all

76 clades (clade-homogeneous). In addition, it is common that all extant taxa are not
77 included in the phylogenetic tree, and the percentage of lineages present is known as the
78 sampling fraction (or ρ) - the ratio of sampled species over the total species diversity for a
79 given clade. By making use of these parameters, a birth-death model allows us to
80 investigate whether the net diversification rate, defined as $r = \lambda - \mu$, has varied over time
81 or among clades (Morlon et al., 2011; Rabosky, 2014; Maliet et al., 2019; Barido-Sottani
82 et al., 2020) and ultimately uncover the processes that have given rise to extant
83 biodiversity in the study group.

84 A SUMMARY OF THE MAIN CONCEPTS AND FINDINGS IN LOUCA AND PENNELL (2020)

85 *The deterministic Lineages-Through-Time plot*

86 The approach to study diversification used by LP relies on the
87 Lineages-Through-Time (LTT) plot (Nee et al., 1992) (Fig. 1), which shows how extant
88 lineages (i.e. only those existing in the present-day) accumulated over time using a
89 phylogenetic tree. Each point in an LTT corresponds to a change in the number of lineages
90 from the root of a phylogenetic tree at $t = 0$ to the present day at $t = T$ (Fig. 1a).
91 Alternatively, as in LP, time can be considered as an age ($\tau = T - t$), where $\tau = 0$ at the
92 present and $\tau = T$ at the origin of the clade, or the root age (Fig. 1b). For consistency
93 with LP, we will generally consider timescale as age (τ) in the equations we use throughout
94 this manuscript.

95 Simply put, when a clade diversifies faster, the slope of the LTT becomes steeper,
96 but when diversification slows, the slope of the LTT levels off. When only extant lineages
97 are considered, as in LP, LTT plots will never exhibit a drop in total lineage diversity over
98 time. Regardless of whether time is τ or t time in the equations, the LTT is usually plotted
99 with the present on the right, thus its slope will never be negative. However, this does not
100 mean that extinction does not have an effect on the shape of the LTT (Nee, 2006). By
101 examining the shape of the LTT plot we can begin to understand how diversification rates

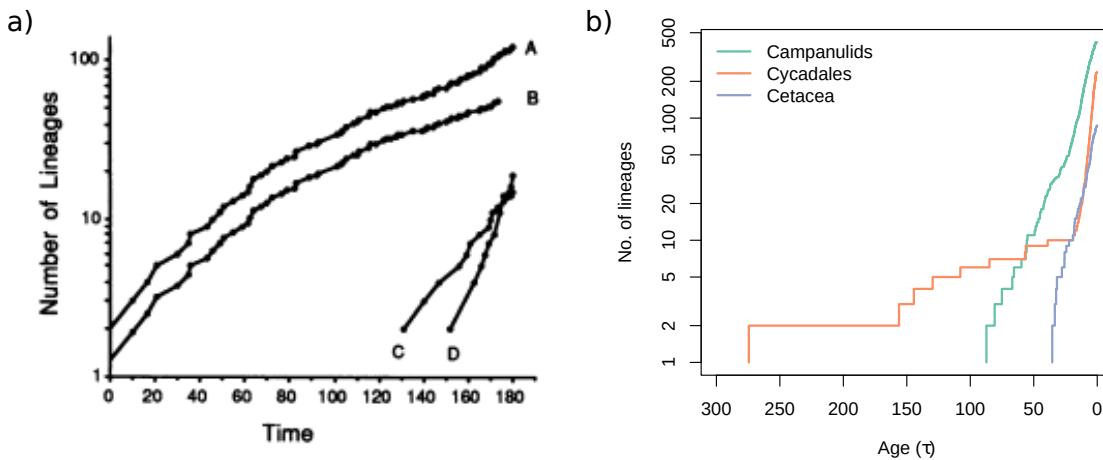


Figure 1. (a) The first example of a lineages-through-time plot (LTU), taken from Nee et al. (1992) and based on a phylogenetic tree of birds. On the y-axis is the number of lineages (log scale) and the x-axis is time since origin (present on the right hand side of the graph). *“Each point corresponds to a change in the number of lineages. Line A, the pattern of origination of all 122 lineages; line B, same as A, but without the Passeri (line C) and the Ciconiiformes (line D). Line B has been shifted downward to aid visual comparison. The diversification rate is quantified by the steepness of the slope.”* In this panel, time is displayed from past to present as time since origin (t). (b) Three LTUs from modern phylogenetic trees of Campanulids (Beaulieu and Donoghue, 2013), Cycadales (Condamine et al., 2015) and Cetacea (Slater et al., 2010). In this panel time is shown from present to past as an age (τ).

102 fluctuated over the history of a clade (Ricklefs, 2007) and develop evolutionary hypotheses
103 on why these fluctuations occurred.

104 To study general properties of phylogenetic trees, a model of the branching process
105 is used. Several models are available, but the birth-death model is the most widely used,
106 and is easily interpreted (Nee, 2006). The birth-death model is a continuous-time Markov
107 chain where at any given age (τ) we can calculate the probability of speciation (birth of a
108 lineage) or extinction (death of a lineage) happening. Because the birth-death process is
109 stochastic, each run (i.e. simulation) will result in a different history of diversification, even
110 if the probabilities for speciation and extinction are the same.

111 For such models we can calculate their expected value, either by averaging over
112 multiple simulations or by approximating it with a set of continuous equations, yielding a
113 deterministic model. Such a model produces the expected value one would get by
114 averaging over an infinite number of simulations, thus it is deterministic because it is fully
115 defined by the parameters, that is, no uncertainty from stochasticity is involved. This

116 latter approach is widely used, and also taken by LP who model the birth-death process as
117 a set of differential equations, which is advantageous because these equations can be solved
118 analytically.

119 LP refer to an LTT generated by such models as a deterministic LTT (dLTT),
120 which corresponds here to the expected LTT generated by trees with given speciation and
121 extinction rates. Empirical LTTs generated using extant timetrees can be compared to
122 model-generated dLTTs (where λ and μ are known) to disentangle how speciation and
123 extinction have influenced patterns of diversity over time. To do this, the probability of the
124 data given the model, or the likelihood, is calculated. Importantly, LP showed that, when
125 λ and μ are clade-homogeneous across the tree, the likelihood can be fully written as a
126 function of the observed LTT and the dLTT (see also Lambert and Stadler (2013)).

127 Typically, by changing the parameters in the model, its dLTT resembles the empirical LTT
128 to a greater or lesser extant, and the model is more or less likely. The best-fitting model
129 can then be selected, representing our best hypothesis for how and to what extent
130 speciation and extinction rates varied over time.

131 *Model congruence and congruence classes*

132 Consider a simple model where λ and μ are fixed over time and among clades and
133 all lineages have been sampled ($\rho = 1$). In this case, the slope of the LTT plot is $r = \lambda - \mu$,
134 except at times close to the present, where the effect of extinction diminishes and the slope
135 becomes λ (Nee et al. (1994); see also Fig. 4 in Nee (2006)). If we know λ we can estimate
136 μ by first estimating the slope of the LTT prior to the upward bend, which corresponds to
137 $r = \lambda - \mu$. This can then be rearranged to $\mu = \lambda - r$ and μ calculated using the estimated
138 values of λ and r (in practice, both parameters can be inferred at once within a likelihood
139 framework using equations in Nee et al. (1994)). LP develop upon this classical knowledge
140 to show that if rates vary over time (τ) it is no longer possible to estimate $\lambda(\tau)$, as the
141 value of $\lambda(\tau)$ at present does not yield any information about its past dynamic. In other

words, it is possible to choose almost any historical scenario for $\lambda(\tau)$ and obtain a complementary scenario of $\mu(\tau)$ that produces the same dLTT. If different models produce the same dLTT then they will also share the same likelihood for any given LTT.

LP call the set of models that generate the same dLTT a "congruence class". These congruence classes contain an infinite number of models with different parameter values that all produce the same dLTT. LP explain that when trying to select the best model we often start with a relatively small set of allowed models that we test. For example, a set of two models where speciation rate is fixed and extinction rate is allowed to vary over time, or *vice versa*. This would produce two equally likely models when trying to explain a slowdown in diversification, one indicating the case was an increase in extinction rate, the other a drop in the speciation rate - there is no way of distinguishing between them Crisp and Cook (2009); Burin et al. (2019). LP suggest that instead of selecting the model closest to the true process, we are instead selecting the model closest to the congruence class that includes the true process (see Fig. 3 in LP). In extreme cases, the best fitting model could thus be further from the true process than a more correct model, just because the former is included in the congruence class and the latter is not. However, LP concede that because we only assess a limited set of models, it is unlikely that we encounter models belonging to the same congruence class, but it is nevertheless possible. The consequence of multiple, equally likely models with different speciation and extinction rates is that these rates cannot be determined. This is a statistical phenomenon known as unidentifiability - the likelihood is the same for multiple parameter values making it impossible to choose one over another.

164 *Unidentifiability*

165 In macroevolutionary modelling we might be interested to know how both λ and μ
166 have changed over time (Alfaro et al., 2009). The unidentifiability issue outlined above
167 means that we would not be able to ascertain the true parameter values of the models that

168 generate our dLTTs. Another well-known example of this in macroevolution is the
169 unidentifiability of α and θ from Ornstein-Uhlenbeck models of trait evolution (Ho and
170 Ané, 2014). This problem is not unique to macroevolutionary models, and, in fact, stems
171 from a basic mathematical issue (Rannala, 2002; Ponciano et al., 2012).

172 Consider a simple example of the concept in which we want to determine the
173 parameter values for x and y . For each value of x in equation 0.1 below, we can find a y
174 that satisfies this equation - and there are an infinite number of equally likely possibilities.
175 It is only when we add more information (in the form of equation 0.2) that we can
176 determine the unique pair of values for x and y . Put simply, a solution can be found only if
177 you possess at least the same number of equations as unknowns. In this case the
178 unidentifiability is caused by overparameterization - there is an excess of parameters such
179 that the model cannot estimate the values of any of them.

$$2x + y = 7 \quad (0.1)$$

180

$$3x + 2y = 12 \quad (0.2)$$

181 Though the LTT is generated through the use of many different observations and
182 elements (DNA, fossils for time-calibration, extant species sampling) it is represented by a
183 single curve made up of one observation at any given point in time that represents the
184 number of lineages in a clade (Fig. 1). Fitting a model to an LTT is like fitting two
185 parameters (a and b) for the slope ($a - b$), which gives you only one value. If we try to
186 estimate a and b separately we find it impossible (Fig. 2a,b). However, we can estimate
187 $a - b$ (Fig. 2c). Estimates of a and b are fully correlated (Fig. 2d) and we find a flat surface
188 in the likelihood where different pairs of values for a and b are equally likely (Fig. 2e),
189 signifying unidentifiability.

190 This problem has been highlighted previously (Nee, 2006), where $a - b$ is the net
191 diversification rate ($r = \lambda - \mu$). However, the birth-death model is more complex than the

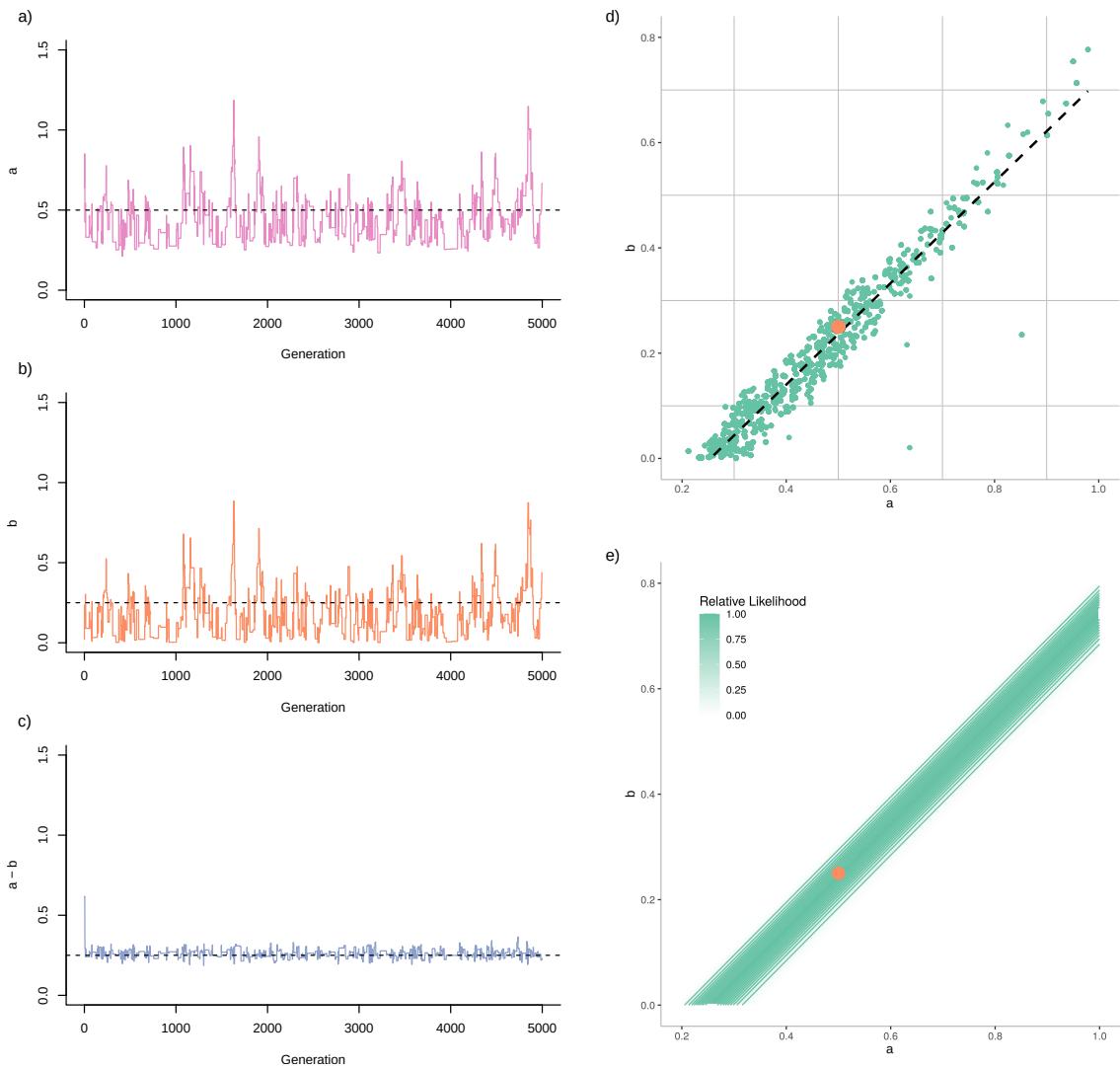


Figure 2. An example of unidentifiability issues encountered when trying to estimate values of two parameters (a & b) for the slope $a - b$. We used a Bayesian Monte-Carlo Markov Chain approach to try to estimate the values of a and b separately. We ran a chain for 5000 generations, sampling each generation. The traces for (a) a and (b) b show a great deal of uncertainty in the parameter estimates compared to the estimates for (c) $a - b$. True values are shown as black dashed lines in panels (a-c) and as orange circles in panels (d-e). We plotted a against b and found that they two parameters were highly correlated (d). When then calculated the relative likelihood over a range of parameters values and found a flat ridge in the likelihood where different pairs of values for a & b are equally likely - or unidentifiable (e).

example illustrated in figure 2. As explained above, speciation and extinction rates are actually identifiable when time-independent because the slope of the LTT reaches λ at the present. Our ability to reliably estimate these traditional diversification rates (λ , μ & r) depends on the amount of information we have available, and the assumptions we make in

	Speciation	Extinction	Sampling fraction	Ref	Identifiable parameters
constant rates	$\lambda = a$	$\mu = b$	$\rho = 1$	Nee et al. (1994)	$r, \lambda, \mu, r_p, \lambda_p, \mu_p$
	$\lambda = a$	$\mu = b$	$\rho < 1$ (known)	Nee et al. (1994)	r, r_p, λ_p, μ_p
	$\lambda = a$	$\mu = b$	$\rho < 1$ (unknown)	Stadler (2013)	r, r_p, λ_p, μ_p
time-dependent rates	$\lambda = f(\tau)$	$\mu = g(\tau)$	$\rho = 1$	Louca and Pennell (2020)	r_p, λ_p, μ_p
	$\lambda = f(\tau)$	$\mu = g(\tau)$	$\rho < 1$ (known)	Louca and Pennell (2020)	r_p, λ_p, μ_p
	$\lambda = f(\tau)$	$\mu = g(\tau)$	$\rho < 1$ (unknown)	Louca and Pennell (2020)	r_p, λ_p, μ_p

Table 1. A table detailing the parameters we can estimate with the Lineages-through-time plot (LTT) approach used in Louca and Pennell (2020) when rates are either constant or time-dependent. When speciation and extinction rate are constant we are able to infer all traditional (r, λ, μ) and pulled rates (r_p, λ_p, μ_p). If sampling fraction (ρ) is unknown, we lose the ability to infer λ and μ . If λ and μ vary over time only pulled rates remain identifiable.

196 our model. For example, if the sampling fraction (ρ) is not known (or assumed) we can no
 197 longer reliably estimate λ and μ because this third unknown parameter in the model leads
 198 to unidentifiability (Table 1). However, $r = \lambda - \mu$ (with ρ) remains identifiable, as the
 199 system is reduced to two parameters only. Likewise, as LP show, if we relax the
 200 assumption of constant rates and allow λ and μ to vary over time, then all traditional
 201 parameters become unidentifiable, including $r(t)$, even if ρ is known or assumed. (Table 1).

202 To exemplify the problem, LP used a very large angiosperm phylogenetic tree
 203 (Smith and Brown, 2018) to show that the observed LTT is congruent with two opposing
 204 scenarios (Fig. 2 in LP): either a continuous increase or a continuous decline in both $\lambda(\tau)$
 205 and $\mu(\tau)$ (though the resulting diversification rates of these two scenarios are very similar).
 206 Therefore, if we observe a rapid increase in the number of lineages in our LTT (Fig. 3) we
 207 cannot determine if it was caused by a decrease in extinction rate, or an increase in
 208 speciation rate. If we want to use models to explain LTTs then traditional variables are
 209 inadequate and we must look towards other possible solutions.

210 *Pulled rates and their interpretation*

211 LP's solution is to use the approach described in Louca et al. (2018), namely not to
 212 estimate $\lambda(\tau)$, $\mu(\tau)$ and ρ , but "pulled" rates that can be directly measured from the shape
 213 of the LTT. There are three pulled rates (λ_p, μ_p, r_p) in Louca et al. (2018). These pulled
 214 rates are based directly on the dLTT - they make use of the slope at a given time (τ) and
 215 the change in the slope, or curvature of the plot. Thus, any dLTT yields a unique set of

216 pulled rates that summarise a congruence class, thereby eliminating the unidentifiability
217 issue. However, these rates are not the speciation and extinction rates everyone knows - so
218 what are they and how are they different from traditional rates?

219 An important consequence of using extant timetrees when investigating patterns of
220 diversification is that LTT plots will likely underestimate the number of lineages at any
221 given time because our trees are missing species (Ricklefs, 2007; Silvestro et al., 2018).
222 Species can be missing for two reasons: (1) they went extinct, or (2) they are extant but
223 were not sampled. However, these two factors will have different effects on the LTT and
224 our estimates of diversification rates. Extinction must have occurred in the past. Lineages
225 that originated recently have had less time to go extinct (Nee et al., 1994; Ricklefs, 2007),
226 so the effect of extinction on our estimates using only extant species is reduced towards the
227 present. As mentioned above, this leads to an increase in the rate of lineage accumulation
228 towards the present, as the effect of extinction decreases, which occurs even when rates are
229 constant (Fig. 3). Conversely, incomplete sampling of a group occurs up to the present day
230 and more strongly affects estimates of recent history (Phillimore and Price, 2008; Heath
231 et al., 2008; Cusimano and Renner, 2010), as the deeper nodes in the phylogeny can be
232 reconstructed with only a few species. The relative importance of extinction and sampling
233 fraction will influence whether λ_p departs from λ more in the past or in the present. To
234 summarise, the presence of extinction will cause us to underestimate speciation rate
235 further in the past, because the number of extinct species increases as we consider more
236 time, while incomplete sampling will lead to underestimates of speciation rates that are
237 more recent (Heath et al., 2008; Cusimano and Renner, 2010; Brock et al., 2011).

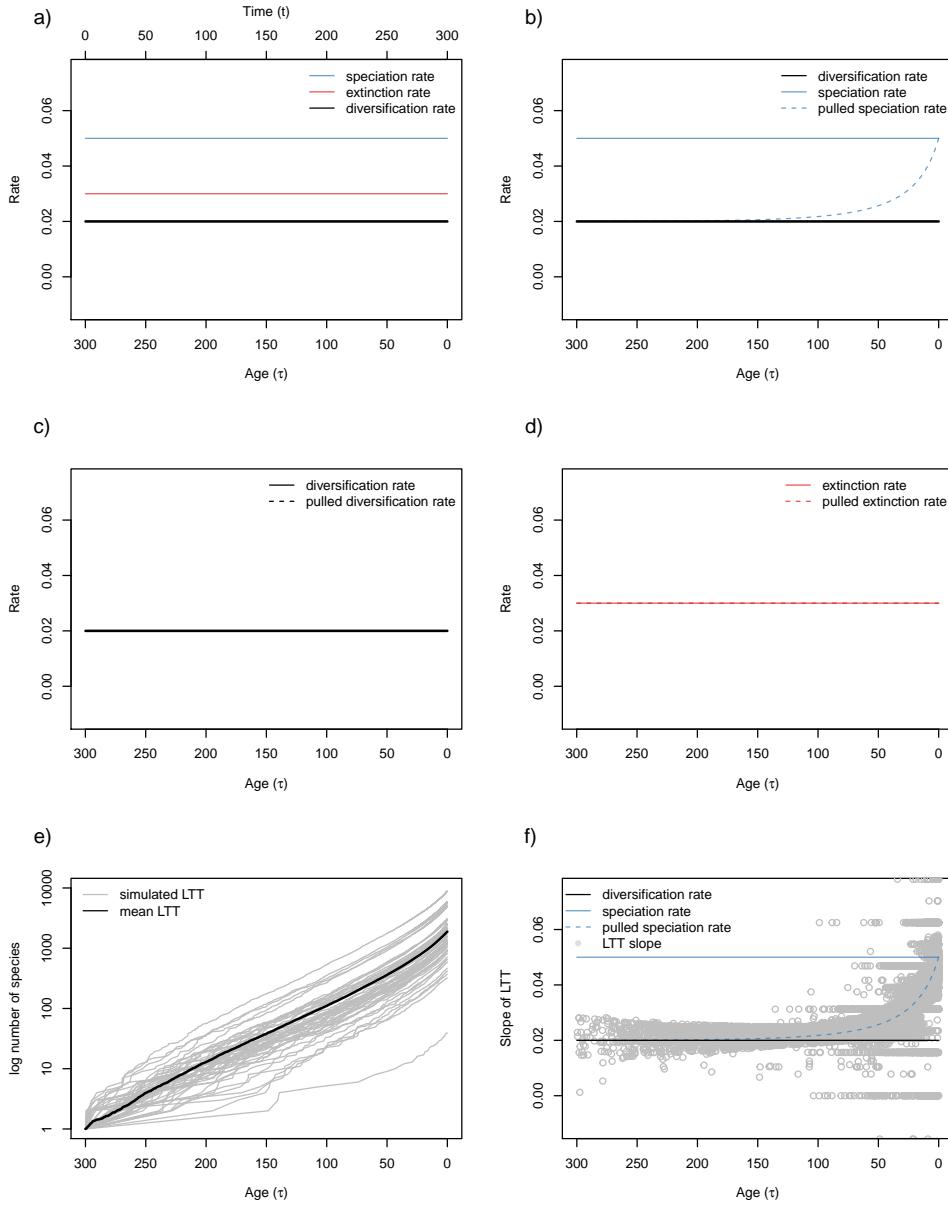


Figure 3. A simple example of the relationship between constant diversification rates and corresponding pulled rates. Panel (a) shows values of speciation rate (λ), extinction rate (μ) and diversification rate (r) over time. An additional axis, at the top of panel (a) shows time since origin (t). Panel (b) shows how in the past, pulled speciation rate (λ_p) is identical to the diversification rate (if sampling fraction = 100%) while closer to the present λ_p approaches speciation rate. The following two panels compare (c) r & pulled diversification rate (r_p) and (d) compares μ & pulled extinction rate (μ_p). In these two cases the pulled rates are identical to the traditional rates. Panel (e) shows 50 LTT plots (grey lines) simulated with the parameters used in panels (a-d) and the mean LTT (black line). Panel (f) shows the slopes of the LTTs in panel (e) over time, matching λ_p and depicting the expected increase towards the present caused by the lack of effect of extinction - lineages do not have enough time to go extinct towards the present. An interactive version of this plot, in which parameters can be modified, can be found at https://ajhelmstetter.shinyapps.io/pulled_rates/.

238 Formally, at a given time (τ), λ_p is the estimated speciation rate multiplied by 1
239 minus the probability that a lineage is missing from the tree due to extinction or
240 incomplete sampling, E . We will not go into details regarding the calculation of E here,
241 but further information can be found in supplementary materials of LP. λ_p is calculated
242 using the following equation:

$$\lambda_p(\tau) = \lambda(\tau)(1 - E(\tau)) \quad (0.3)$$

243 So, if all species are in the tree and there is no extinction (i.e. the probability of
244 missing lineages, or E , is 0) then the λ_p is equal to the (un-pulled) speciation rate, λ . Any
245 increase in extinction rate or the number of unsampled lineages (i.e. $E > 0$) will cause λ_p
246 to drop, or be "pulled", below speciation rate (Fig. 3). The lower the extinction rate and
247 the greater the sampling fraction, the closer the estimate of λ_p will be to λ .

248 LP also use pulled diversification rate (r_p), a parameter that is similar to the net
249 diversification rate ($r = \lambda - \mu$) but is again modified by another term. This new term is
250 the relative ($\frac{1}{\lambda}$) rate of change in speciation rate over time ($\frac{d\lambda}{d\tau}$). This causes the pulled
251 diversification rate to lag behind the unpulled rate. The "pull" of r_p is actually a delay in
252 the response of this parameter when compared to diversification rate. This is in contrast to
253 the "pull" of λ_p , which refers to a reduction in the estimated value of λ_p relative to λ .
254 Pulled diversification rate can be represented by the following equation :

$$r_p = \lambda - \mu + \frac{1}{\lambda} \cdot \frac{d\lambda}{d\tau} \quad (0.4)$$

255 Consider an example where we have an increase in speciation rate at around 100
256 Ma in a clade (Fig. 4). When considering time as an age (using τ), speciation rate
257 increases as τ decreases from the origin of the clade ($\tau = 300$ Ma) to the present ($\tau = 0$).
258 This means that when speciation accelerates, $\frac{d\lambda}{d\tau}$ is negative. This leads to a "drop" in r_p
259 (Fig. 4c) before it stabilizes at a new value of r_p that is higher than the previous value,
260 reflecting the increase in λ . However, the change in the slope of the LTT (Fig. 4e,f) is

261 minimal, so this is not actually measurable from a phylogenetic tree. We note that LP also
262 defined a pulled extinction rate, (μ_p), which is similar to r_p so we do not discuss it here
263 (see LP, Louca et al. (2018) for further details).

264 The difference between the true diversification rate and an estimated r_p can be
265 likened to a race between an amateur and a professional race car driver. The professional
266 driver, representing the true diversification rate in our analogy, hits the apex of each
267 corner, going smoothly around a racetrack until the finish line. The amateur, representing
268 r_p , will eventually arrive at the finish line, but may exceed track limits a few times when
269 doing so because of their poor reactions. However, if the track is simply a straight line
270 both will perform equally well. This is because the r_p is equal to the diversification rate
271 ($r = \lambda - \mu$) whenever λ is constant in time ($\frac{d\lambda}{d\tau} = 0$), but differs from r when λ varies with
272 time (see Technical considerations below for more details).

273 With these new variables we can revisit questions such as: has diversification been
274 constant over time? Pulled rates can be estimated with many commonly used models of
275 diversification (Louca and Pennell, 2020). For example, λ_p is the speciation rate one would
276 get by constraining extinction to be 0 and assuming complete species sampling. For r_p this
277 involves estimating r by making λ time-independent. In summary, λ_p provides information
278 about how λ changes over time while taking into account past extinction and the
279 proportion of lineages sampled. r_p provides a slightly delayed estimate of r with extreme
280 responses to rapid changes in λ . While λ_p can be very different from the underlying
281 speciation and extinction rates, r_p is close to the net diversification rate as long as λ does
282 not change too rapidly.

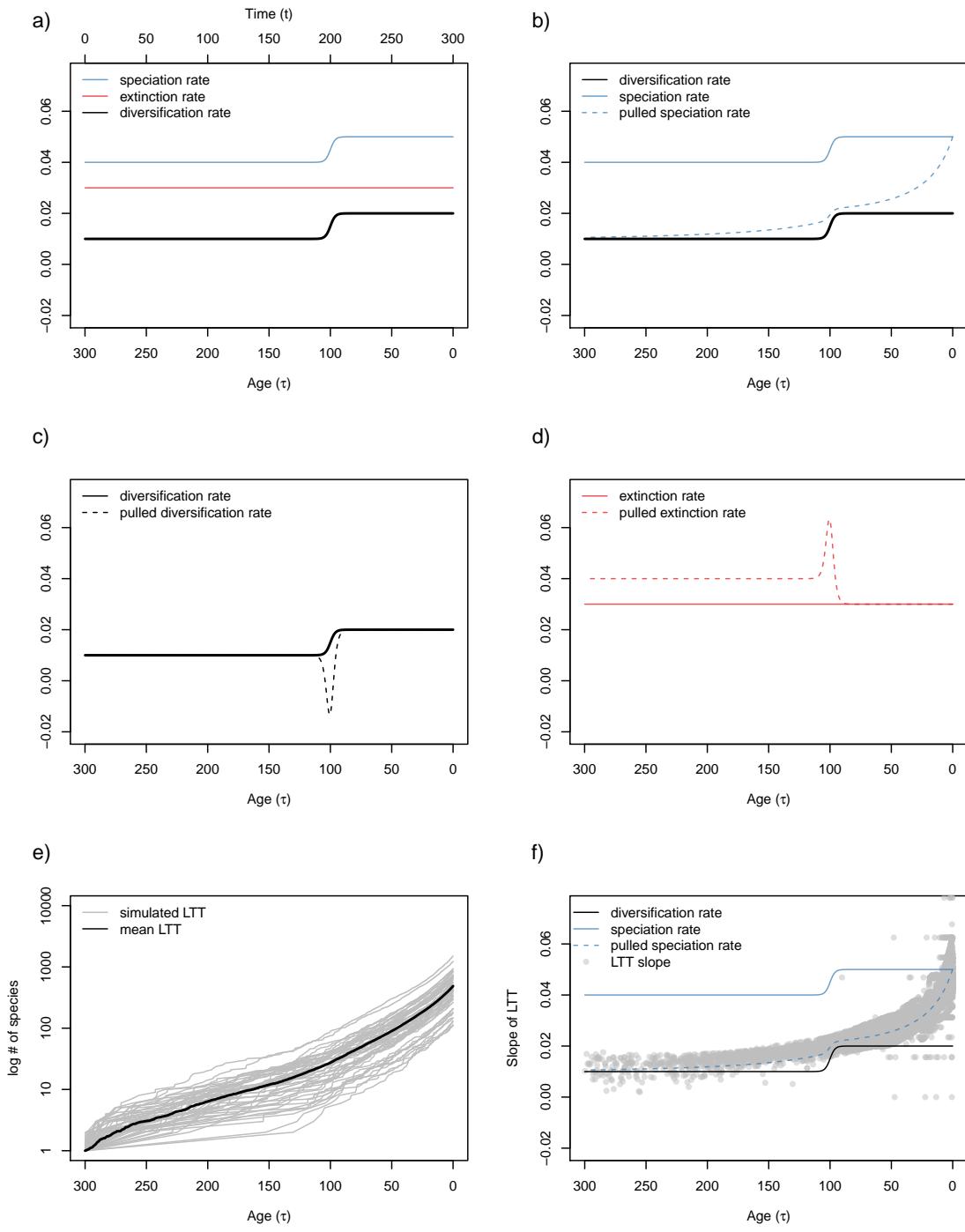


Figure 4. A slightly more complex example of the relationship between diversification rates and corresponding pulled rates where a single shift - an increase in speciation rate - has taken place. Panel (a) shows values of speciation rate (λ), extinction rate (μ) and diversification rate (r) over time. An additional axis, at the top of panel (a) shows time since origin (t). Panel (b) shows the gradual change in pulled speciation rate (λ_p) during the shift in λ . Panel (c) compares r and pulled diversification rate (r_p). The sudden increase in λ causes r_p to decrease suddenly before recovering to the r . Panel (d) compares μ and pulled extinction rate (μ_p) and shows an inverse pattern to panel (c). Panel (e) shows 50 LTT plots (grey lines) simulated with the parameters used in panels (a-d) and the mean LTT (black line). Panel (f) shows the slopes of the LTTs in panel (e) over time, matching λ_p and again depicting the expected increase towards the present caused by the lack of effect of extinction.

283

TECHNICAL CONSIDERATIONS

284

How continuously can speciation and extinction rates vary?

285

Although this is standard practice, it should be noted that the approach of LP considers speciation and extinction to be continuous processes: at any infinitesimal time interval, the species number changes infinitesimally through speciation and extinction. In the birth-death process, however, the smallest amount of change in the number of species is one, and this happens only at particular moments in time. Even if speciation, in reality, is a complex process that takes time (Etienne and Rosindell, 2012), it is sufficiently fast to be considered instantaneous in evolutionary time. An empirical LTT plot will thus show discrete events, rather than being a continuous function, as is the dLTT. To measure the pulled rates, LP propose calculating the slope and curvature of the LTT plot. For the dLTT, where the number of lineages is a continuous function of time, these are the first and second derivative of this function. For empirical LTT plots, one has to calculate the slope and curvature using some time interval. When working with a large phylogenetic tree and many species (as in the examples discussed by LP and (Louca et al., 2018)), the LTT is smooth and the slope and curvature, which are necessary for the estimation of the pulled rates, can be reliably estimated. However, many studies attempt to estimate diversification rates with relatively small numbers of species (i.e. < 1000 (Hutter et al., 2017) or even < 100 Duan et al. (2018)). Thus, as the number of species diminishes, one has to be aware that overparameterization might occur, and it would be wiser to stick to simple functions of diversification (or their pulled variants) through time, the simplest being time-independent rates. Furthermore, even in large trees, rates will always be estimated using a time interval that contains a sufficient number of speciation and extinction events. The consequence of this is that rapid changes in diversification rates might be missed due to the resolution of the chosen interval.

308 *The "pull" in r_p is a result of the lag time between extinction and speciation*

309 Consider a simple case with no extinction ($\mu = 0$) so that changes in r that only
310 come from changes in λ . If so, $r = \lambda$ but r_p is not exactly λ because of temporal variations
311 in λ (the term $\frac{1}{\lambda} \cdot \frac{d\lambda}{d\tau}$ in equation (0.4)). LP suggest that "*the pulled diversification rate can*
312 *be interpreted as the effective net diversification rate if λ was time-independent*". In our
313 example, this means replacing a scenario where μ is constant (at 0) and λ varies with a
314 scenario where λ is constant and μ varies, as in LP. The difficulty of using μ to explain
315 variation is that there is a slight delay between the effect of speciation and the effect of
316 extinction. It is necessary to wait for species to arise before they can go extinct.

317 As mentioned previously, lineages that originated more recently have had less time
318 to go extinct. In a constant birth-death process, this is only visible in recent history: the
319 slope of the LTT is $r = \lambda - \mu$ during most of the past but increases to λ for very recent
320 times where the stationary behaviour has not yet been reached (Fig. 3). However, this
321 phenomenon is not unique to very recent times - it will also occur whenever there is a
322 change in speciation rate. Ultimately, this is the cause of the difference between r_p and r .

323 To illustrate this, imagine a massive increase in the number of lineages caused by a
324 burst of speciation (Fig. 4). Over a short time period many new lineages have become
325 available for potential extinction but they have yet to go extinct because not enough time
326 has passed since they appeared for extinction to take place. There is now a disequilibrium
327 between speciation and extinction, manifested as a lag in the time extinction takes to
328 affect all of the new lineages. As time continues, these numerous new lineages will begin to
329 go extinct, meaning that frequency of species extinction will increase to "catch up" to
330 speciation and reach a new stationary point. This effect is stronger when λ varies rapidly
331 (i.e. high $\frac{1}{\lambda} \cdot \frac{d\lambda}{d\tau}$). Conversely, speciation cannot occur in a lineage after it has gone extinct,
332 so there is no similar lag caused by changes in extinction rate. This is also why variation in
333 extinction rate would not cause r_p to deviate from r (Fig. 5b).

334

DISCUSSION

335 A recent study by Morlon et al. (2020) presents an alternative point of view that
336 opposes the conclusions in LP. They focus on how a hypothesis-based framework allows us
337 to overcome many of the issues that are raised in LP. Indeed, we are limiting our set of
338 models to be tested to only those that represent our hypotheses about the factors shaping
339 diversification in a given group. We are not often interested in determining the precise
340 values of speciation and extinction rate but rather how different diversification scenarios
341 summarised by models containing λ and μ explain patterns in a phylogenetic tree. The
342 criticisms put forward by Morlon et al. (2020) will stimulate important discussion of
343 considerations when using diversification models. We extend this discussion by highlighting
344 several key points that must be considered in addition.

345

Uses and limitations of LTTs

346 LTTs are a simplistic way to visualize and summarize a time-calibrated
347 phylogenetic tree, ignoring the full distribution of branch lengths, tree topology and
348 extinct species (Morlon et al., 2011). However, under the assumption that λ and μ are
349 clade-homogeneous, LP showed that the LTT contains the complete information about the
350 underlying branching process (See also Lambert and Stadler (2013)). This simplicity
351 provided the opportunity for LP to show mathematically how LTTs can lead to
352 misinterpretation. These issues are not new to macroevolutionary biology. A review by Nee
353 (2006) clearly demonstrated how an LTT may change when extinction is present alongside
354 speciation (birth-death), as opposed to speciation alone (pure-birth), summarising theory
355 from previous works (Nee et al., 1992, 1994; Harvey et al., 1994). If the growth of an
356 extant timetree is represented as an LTT on a semi-log scale (i.e. lineage number is
357 logarithmic, time is not, see Fig. 1) we would expect the trend to be linear under a
358 pure-birth process (with constant speciation and no extinction). If extinction is introduced
359 then the LTT would deviate from this linearity. When both rates are constant and greater

360 than 0, the curve is expected to be linear over most of its history, but as time reaches the
361 present the rate of lineage accumulation will increase (i.e. the LTT slope will become
362 steeper), as shown in Figure 3a. With no prior knowledge of the parameters, this could be,
363 at least in part, because of increasing speciation rate towards the present (Fig. 3b), instead
364 of decreasing effect of extinction (Fig. 3). It is important to keep in mind that we are
365 dealing with a phylogenetic tree made up of entirely extant species. The unobserved
366 branches of species that went extinct (and are therefore not in the extant timetree) do not
367 contribute to the lineage counts in the LTT, making the estimated lineage accumulation
368 rate lower in the past (or "pulling" it down). Nee et al. (1994) highlighted this issue 20
369 years ago in the context of models where diversification rates were constant over time and
370 now LP have provided an important extension of this idea to models that allow for rates to
371 vary through time. Since Nee et al. (1994), the well-known limitations of LTTs for inferring
372 speciation and extinction rates have continued to be addressed in other studies (Ricklefs,
373 2007; Vamosi et al., 2018; Rabosky and Lovette, 2008; Crisp and Cook, 2009), and most
374 recently in LP.

375

Diversification rates vary among clades

376 The conclusions of LP imply that we can test hypotheses about whether
377 diversification rates deviate from constancy over time using pulled rates. We would be
378 unable to pin this on changes in speciation or extinction rates, but would get a sense of
379 how variable diversification has been (Burin et al., 2019). This would be useful for testing
380 whether diversification in particular clades has remained constant or been subject to large
381 shifts in diversification (e.g. mass extinctions) but not when diversification rate has shifted
382 in a subclade (e.g. due to the evolution of a key innovation). The first use of pulled rates
383 was in Louca et al. (2018), where they studied bacterial diversification, stating "*Our*
384 *findings suggest that, during the past 1 billion years, global bacterial speciation and*
385 *extinction rates were not substantially affected during the mass extinction events seen in*

386 *eukaryotic fossil records.*" This might suggest that nothing particularly extraordinary
387 happened in the macroevolutionary dynamics of bacteria in the last billion years. However,
388 it is important to note that the models used in Louca et al. (2018) and in Louca and
389 Pennell (2020) do not allow rates to vary among clades. The rates estimated using such
390 clade-homogeneous models will correspond to the average rates over time in the entire
391 study group, therefore missing out on any variation among clades - for example any
392 difference in diversification rates between those species that use terrestrial *versus* marine
393 environments (Louca et al., 2018). Subclades are important in driving inferred
394 diversification patterns (see Morlon et al. (2011); Maliet et al. (2019); Rabosky (2020)), so
395 this may mean that we miss out on influential and interesting dynamics when using pulled
396 rates. Louca et al. (2018) touch on this point themselves: "*It is possible that diversification*
397 *within individual bacterial clades may have been influenced by eukaryotic radiations and*
398 *extinctions, and that these cases are overshadowed when considering all bacteria together.*"
399 Given the diversity of life on Earth, it is unrealistic to assume that major events would
400 have had the same effect on all lineages of a large, cosmopolitan clade, with vast amounts
401 of genetic, morphological and ecological variation. Such an assumption prevents us from
402 investigating some of the most interesting and fundamental questions in macroevolutionary
403 biology e.g. why are some clades more diverse than others?

404 The same criticisms could be levelled at LP's use of a large phylogenetic tree of
405 angiosperms (Smith and Brown, 2018) that contains more than 65,000 of the roughly
406 300,000 known species, ranging from small ephemeral plants like *Arabidopsis thaliana* to
407 gigantic, long-lived trees such as *Eucalyptus regnans*. Furthermore, a large amount of
408 research has shown that diversification rates have varied significantly among
409 flowering-plant clades (e.g. O'Meara et al. (2016); Igea et al. (2017); Vamosi et al. (2018);
410 Onstein (2019); Soltis et al. (2019); Zenil-Ferguson et al. (2019); Magallón et al. (2019)),
411 and pulled rates would not be able to contribute to furthering our understanding of why
412 this may be.

413 Fortunately, the assumption of homogeneous rates among clades is not common in
414 modern approaches. For instance, Bayesian Analysis of Macroevolutionary Mixtures
415 (BAMM) (Rabosky, 2014) is one of several methods (Alfaro et al., 2009; Morlon et al.,
416 2011; Höhna et al., 2016a; Maliet et al., 2019; Barido-Sottani et al., 2020) that relaxes the
417 assumption that all lineages share the same evolutionary rates at a given point in time
418 (Rabosky, 2017). This is a key difference from the models used by LP because it allows
419 lineages to differ in their rates of speciation and extinction. With BAMM, the entire
420 phylogeny could be described using a model similar to what is used in Louca and Pennell
421 (2020), or alternatively, it could be described using multiple processes that explain rates of
422 diversification on different parts of the tree. Within each of these processes, λ and μ
423 probably still faces the same unidentifiable issues outlined in LP. However, BAMM makes
424 use of the full topology that includes information (e.g. branch lengths) that the LTT lacks.
425 A model that allows diversification rate varies among clades could yield entirely different
426 insights compared to a model where this rate can only vary in time.

427 Another model commonly used to estimate and compare diversification rates among
428 clades is the Binary-State Speciation and Extinction (BiSSE) model (Maddison et al.,
429 2007), part of a family of models known as the state-dependent models of diversification
430 (-SSE models (Ng and Smith, 2014; O'Meara and Beaulieu, 2016; Beaulieu and O'Meara,
431 2016; Caetano et al., 2018)). These models are extensions of the birth-death model that
432 also includes information about character states of extant species. SSE models jointly
433 estimate ancestral states at each node of the phylogenetic tree, rates of transition between
434 character states, and state-dependent diversification rates. LP state that the likelihood
435 functions of SSE models are too complex to be addressed in their study, but suggest that
436 the same problems they uncover probably still apply. As in BAMM, BiSSE makes use of
437 the full tree topology (Maddison et al., 2007) - it also considers character state evolution,
438 rather than just the timing of branching events as in the LTT (Nee et al., 1994). LP
439 further suggest that it remains unclear how the dependence on character states (which, if

440 removed, collapses equations in BiSSE to those shown in Nee et al. (1994)) affects the
441 unidentifiability issue they raise. In the original BiSSE model (Maddison et al., 2007), two
442 important and relevant assumptions were made: (1) sampling fraction is assumed to be
443 100%, and (2) speciation, extinction and transition rate remain constant through time per
444 character state.

445 These may allow the BiSSE model to overcome (or pre-empt) some of the problems
446 raised by LP. LP show that λ equals λ_p when sampling fraction is 100% and $\mu = 0$. The
447 first of these was assumed in the original BiSSE model, though it has since been relaxed
448 (FitzJohn et al., 2009). Extinction can easily be set to 0 in these models, which satisfies
449 the second BiSSE assumption and allows estimation of λ (e.g. Joly and Schoen (2021)).
450 Similarly, r_p equals r when λ is constant, also an assumption in BiSSE. With these
451 additional assumptions, the congruence class collapses to only one model. However, this
452 does not mean that BiSSE actually estimates the speciation rate of a clade correctly; LP's
453 results show we should take this BiSSE estimate as a proxy for the diversification rate.
454 Nevertheless, we stress that the likelihood of time-dependent diversification models (as in
455 LP) is not the same as the likelihood of state-dependent diversification models (-SSE
456 models) and what is unidentifiable in the former does not tell us about identifiability in
457 the latter.

458 It is unclear how clade-heterogeneous rates would affect model congruence, and how
459 the additional information included when using models included in programs such as
460 BiSSE and BAMM would affect the unidentifiability issues. However, what is clearer is that
461 the issues raised in LP cannot be readily applied to commonly used macroevolutionary
462 approaches without further work to show that criticisms related to LTT-based approaches
463 are applicable to these more complex models. Even if unidentifiability issues remain in
464 such models they may not be relevant to the questions the models were built to answer, for
465 example those models that test for variation in diversification rates in association with
466 particular clades or traits. In cases like these, it is not the precise values of rates that are

467 important but instead whether rates in one group of lineages are higher than another.

468 Perhaps most importantly, this means that we should not forego building models
469 that estimate diversification rates because one, simplistic approach has problems known
470 from a long time, but instead continue to improve them and build upon the work done in
471 LP. A case in point is the issue of null model choice when using SSE models raised by
472 Rabosky and Goldberg (2015). This criticism spurred on innovation that led to the
473 development of models with hidden states (Beaulieu and O'Meara, 2016), which are now
474 present in various new incarnations of the SSE approach (e.g. (Caetano et al., 2018;
475 Herrera-Alsina et al., 2019)).

476 *Pulled rates are difficult to interpret*

477 LP compared the usefulness of pulled rates to effective population size (N_e) in
478 population genetics. N_e can be broadly defined as the number of breeding individuals in an
479 idealised population (e.g. constant size, random mating etc.) that would be able to explain
480 the summary statistics of an observed population (e.g. amount of polymorphism, level of
481 inbreeding). N_e is fairly intuitive and will react to biological phenomena in expected ways
482 (e.g. under population structure (Whitlock and Barton, 1997) or non-random mating
483 (Caballero and Hill, 1992)).

484 LP state that the variables they introduce are "easily interpretable". Their
485 terminology, however, is not completely consistent nor coherent with more traditional uses,
486 which can cause confusion. Given that $r = \lambda - \mu$ one might intuitively think that
487 $r_p = \lambda_p - \mu_p$ but this is not the case. Pulled rates are simply different ways of summarizing
488 congruence classes and each one is calculated using both speciation and extinction rates.
489 λ_p is reasonably intuitive, though given that extinction is also included it is more similar to
490 a diversification rate than a speciation rate. Indeed, λ_p is defined as the slope of the LTT
491 plot (Louca et al., 2018) (see Fig. 3f, 4f), which corresponds to the past diversification rate,
492 and to the speciation rate at present in the case all extant species are included (Nee et al.,

493 1994).

494 Pulled diversification rate, however, is much more difficult to interpret, perhaps
495 initially because the "pull" of r_p is not the same as the "pull" of λ_p . Whereas λ_p decreases
496 in value relative to λ , r_p is delayed in time relative to r (Fig. 5) and could better be termed
497 as "delayed" rather than "pulled". We simulated a variety of diversification scenarios from
498 simple to more complex (Fig. 5) and show that r_p and r are similar in each case. However,
499 r_p is not as intuitive as r or N_e . For example, drastic increases in r can lead to sharp
500 decreases in r_p (Fig. 5a). The inverted pattern r_p presents in this case would make it
501 challenging to present in a clear and concise way. Given the added difficulty of its
502 interpretation we question whether r_p provides a more useful estimate of the process of
503 diversification than an estimate of r .

504 However, compared to other pulled rates, r_p could be particularly useful, not as an
505 effective parameter, like N_e , but as a reasonable approximation of the true r . Indeed, we
506 noted above that when shifts in λ are not too severe nor too rapid, r_p is close to r (Fig. 5).
507 Trying to biologically interpret fine-scale variations in r_p would certainly lead to spurious
508 conclusions. However, changes in r_p at a large scale are good proxies for large scale
509 variation in r . This is clearly illustrated in figure 5a where the main trend of the r_p is a
510 recent increase in diversification, and in figure 5d where the main trend is the stability of
511 diversification despite rapid, short-term oscillations.

512 Pulled rates can be estimated using only the shape of the LTT plot, without any
513 further information since they are non-parametric estimates that do not suffer from the
514 unidentifiability problems outlined previously. However, they cannot be directly
515 interpreted in biologically meaningful terms. To estimate rates that are meaningful (e.g. λ ,
516 μ and r), one needs to make further assumptions such as constant rates of speciation and
517 extinction over time.

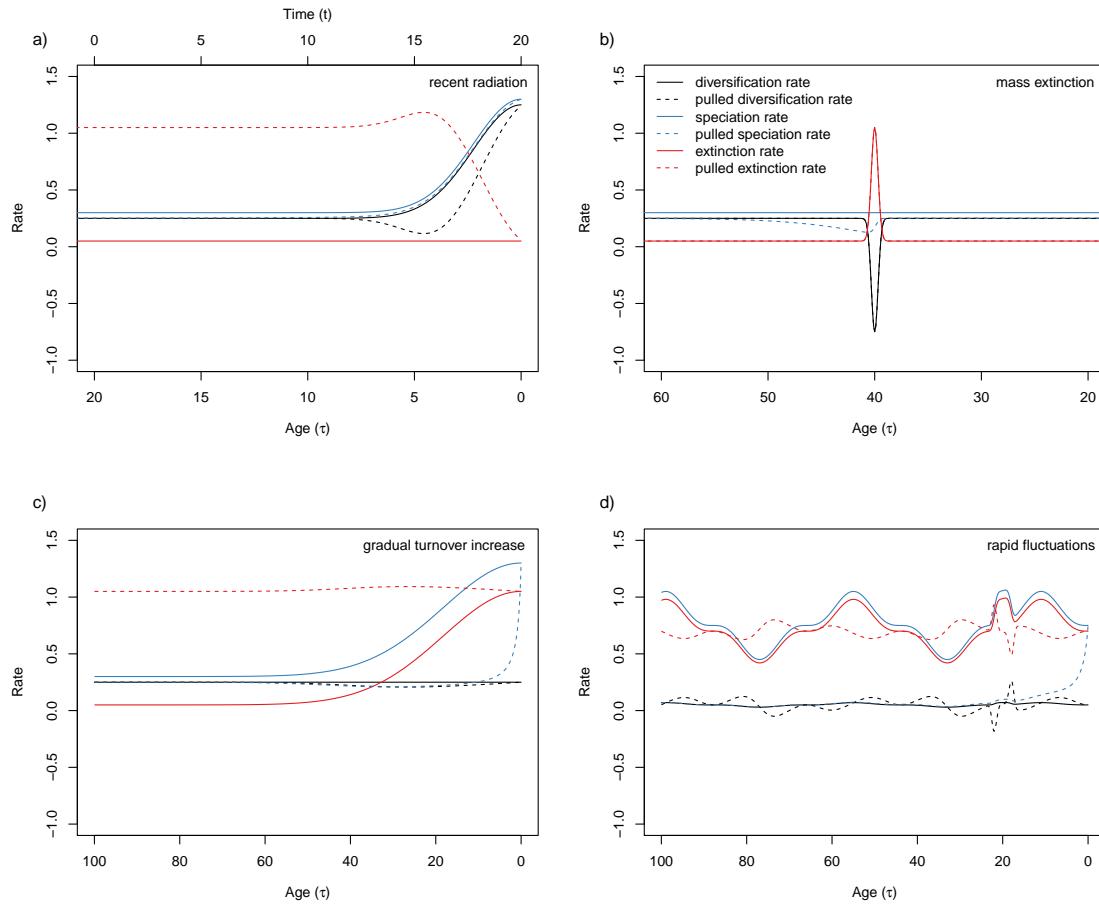


Figure 5. Comparison of traditional and pulled rates under three simulated diversification scenarios that are commonly investigated (a-c) and a final, more complex scenario. Panel (a) shows a recent radiation where diversification rate and speciation rate sharply increase towards the present. An additional axis, at the top of panel (a) shows time going forward (t). Panel (b) shows a mass extinction event at 40 Ma in which extinction rate briefly but rapidly increases and then falls back to previous levels. Panel (c) shows a gradual increase in species turnover rate (both speciation and extinction rate increase slowly over time). Panel (d) shows a scenario where speciation and extinction rates are similar to each other but are in rapid fluctuation over time. This results in a relatively constant diversification rate (r) and a rapidly fluctuating pulled diversification rate (r_p) that remains close to r .

518

On the use of models

519

The discussion sparked by Louca and Pennell (2020) highlights an important issue:

520

evolutionary biologists should be interested in the actual history of diversification of the 521 clades they study. The framework developed by Louca et al. (2018) shows how to do this 522 using the shape of the LTT plot, without making strong assumptions about past speciation 523 and extinction rates. Indeed, the slope and curvature of the LTT plot contain information 524 about the diversification history of the clade. Much of the debate, however, focuses on the

ability to recover a "true" history of diversification. Indeed, the goal of a scientific study should be to find out what really happened, but it becomes confusing if one considers a simulated birth-death process as the "true" history. This birth-death process is determined by two parameters (λ and μ) that can vary over time. These parameters are supposed to correspond to the rate that a lineage splits into two lineages, or goes extinct. In reality, however, a species does not have a speciation and an extinction rate in the same way it has a geographic distribution and a population size. These rates only make sense when they are aggregated over a number of species and a certain amount of evolutionary time. That is, they are descriptive statistics summarizing much more complex processes that are acting at the microevolutionary level, and that would eventually lead to speciation or extinction. Louca and Pennell (2020) convincingly show that one cannot estimate these statistics reliably from LTT plots, and propose statistics that can be estimated more reliably. That these alternative statistics do not exactly correspond to the parameters of the birth-death process is not a problem; the birth-death process is only a model of diversification, and not the truth about diversification itself. The framework built by Louca et al. (2018) and LP allows us to use the LTT to test whether the diversification rate was constant or not. If researchers want to know how speciation and extinction actually changed to give rise to this diversification history, they will have to use other methods.

CONCLUSION

Louca and Pennell (2020) have pointed out key issues with how we approach macroevolutionary modelling, namely the inability to distinguish historical diversification scenarios under certain circumstances. Their formalization of the unidentifiability issues in LTT-based models is an important step forward that provides the mathematical tools to study the associated issues further. LP highlight the avenues we must consider and develop upon to ensure we do not make similar mistakes in the future. Whether variation in diversification rate is due to changes in speciation or extinction is certainly an interesting

avenue of research, but LP have shown that exploring this would require much more than just fitting a model with speciation and extinction rates to an LTT. Indeed, more recent diversification models go beyond this by making use of additional information that is ignored by models relying only on the LTT. Awareness and consideration of potential unidentifiability issues is important for macroevolutionary biologists going forward when they employ such models of diversification. However, it is important to note that LP does not show that speciation and extinction cannot be estimated with evolutionary trees (Pagel, 2020). Instead, they show that when using extant timetrees with a single LTT-based approach, unidentifiability issues are encountered in the estimation of speciation and extinction rates, and that these problems can be circumvented by making use of pulled rates. Further work is needed to identify the extent to which the issues raised in LP apply to the more complex models of diversification used today. Comparisons should be made in empirical studies that use both traditional and pulled rates, to see if differences in results exist between these approaches in practice. In the meantime it is important that the field continues to grow by using and building upon modern macroevolutionary methods, albeit with a critical eye.

567

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579

SUPPLEMENTARY MATERIAL

580

Code associated with this manuscript is available from

581 http://github.com/ajhelmstetter/pulled_rates

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791

FIGURE LEGENDS

792 **Figure 1** (a) The first example of a lineages-through-time plot (LTT), taken from
793 Nee et al. (1992) and based on a phylogenetic tree of birds. On the y-axis is the number of
794 lineages (log scale) and the x-axis is time since origin (present on the right hand side of the
795 graph). *"Each point corresponds to a change in the number of lineages. Line A, the pattern*
796 *of origination of all 122 lineages; line B, same as A, but without the Passeri (line C) and*
797 *the Ciconiiformes (line D). Line B has been shifted downward to aid visual comparison.*

798 *The diversification rate is quantified by the steepness of the slope."* In this panel, time is
799 displayed from past to present as time since origin (t). (b) Three LTTs from modern
800 phylogenetic trees of Campanulids (Beaulieu and Donoghue, 2013), Cycadales (Condamine
801 et al., 2015) and Cetacea (Slater et al., 2010). In this panel time is shown from present to
802 past as an age (τ).

803

Figure 2 An example of unidentifiability issues encountered when trying to
804 estimate values of two parameters (a & b) for the slope $a - b$. We used a Bayesian
805 Monte-Carlo Markov Chain approach to try to estimate the values of a and b separately.
806 We ran a chain for 5000 generations, sampling each generation. The traces for (a) a and
807 (b) b show a great deal of uncertainty in the parameter estimates compared to the
808 estimates for (c) $a - b$. True values are shown as black dashed lines in panels (a-c) and as
809 orange circles in panels (d-e). We plotted a against b and found that they two parameters
810 were highly correlated (d). When then calculated the relative likelihood over a range of
811 parameters values and found a flat ridge in the likelihood where different pairs of values for
812 a & b are equally likely - or unidentifiable (e).

813

Figure 3 A simple example of the relationship between constant diversification
814 rates and corresponding pulled rates. Panel (a) shows values of speciation rate (λ),
815 extinction rate (μ) and diversification rate (r) over time. An additional axis, at the top of
816 panel (a) shows time since origin (t). Panel (b) shows how in the past, pulled speciation

rate (λ_p) is identical to the diversification rate (if sampling fraction = 100%) while closer to the present λ_p approaches speciation rate. The following two panels compare (c) r & pulled diversification rate (r_p) and (d) compares μ & pulled extinction rate (μ_p). In these two cases the pulled rates are identical to the traditional rates. Panel (e) shows 50 LTT plots (grey lines) simulated with the parameters used in panels (a-d) and the mean LTT (black line). Panel (f) shows the slopes of the LTTs in panel (e) over time, matching λ_p and depicting the expected increase towards the present caused by the lack of effect of extinction - lineages do not have enough time to go extinct towards the present. An interactive version of this plot, in which parameters can be modified, can be found at https://ajhelmstetter.shinyapps.io/pulled_rates/.

Figure 4 A slightly more complex example of the relationship between diversification rates and corresponding pulled rates where a single shift - an increase in speciation rate - has taken place. Panel (a) shows values of speciation rate (λ), extinction rate (μ) and diversification rate (r) over time. An additional axis, at the top of panel (a) shows time since origin (t). Panel (b) shows the gradual change in pulled speciation rate (λ_p) during the shift in λ . Panel (c) compares r and pulled diversification rate (r_p). The sudden increase in λ causes r_p to decrease suddenly before recovering to the r . Panel (d) compares μ and pulled extinction rate (μ_p) and shows an inverse pattern to panel (c). Panel (e) shows 50 LTT plots (grey lines) simulated with the parameters used in panels (a-d) and the mean LTT (black line). Panel (f) shows the slopes of the LTTs in panel (e) over time, matching λ_p and again depicting the expected increase towards the present caused by the lack of effect of extinction.

Figure 5 Comparison of traditional and pulled rates under three simulated diversification scenarios that are commonly investigated (a-c) and a final, more complex scenario. Panel (a) shows a recent radiation where diversification rate and speciation rate sharply increase towards the present. An additional axis, at the top of panel (a) shows time going forward (t). Panel (b) shows a mass extinction event at 40 Ma in which

844 extinction rate briefly but rapidly increases and then falls back to previous levels. Panel (c)
845 shows a gradual increase in species turnover rate (both speciation and extinction rate
846 increase slowly over time). Panel (d) shows a scenario where speciation and extinction
847 rates are similar to each other but are in rapid fluctuation over time. This results in a
848 relatively constant diversification rate (r) and a rapidly fluctuating pulled diversification
849 rate (r_p) that remains close to r .