

1 The error in Bayesian phylogenetic reconstruction  
2 when speciation co-occurs

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8 **Abstract**

9 There exist millions of species on Earth, all originating from a common  
10 ancestor billions of years ago. The field of phylogenetics uses heritable  
11 material (e.g. DNA) to determine the evolutionary history of species.

12 Starting from heritable material and explicit assumptions, Bayesian  
13 phylogenetics allows to infer a jointly-estimated phylogeny and parameter  
14 estimates distribution. One of these assumptions in the speciation model,  
15 which mathematically describes the branching process of a phylogeny in  
16 time. The most used speciation model assumes that speciation events are  
17 independent, where we know that certain events can trigger speciation  
18 events in multiple species.

19 This research answers the question what the impact is of using a species  
20 tree model that assumes speciation is independent, when it is used on  
21 phylogenies created by a tree model in which speciation can co-occur.

Here we show the inference error made, when nature has varying degrees of co-occurring speciation over a wide range of parameter settings.

We show that the inference error correlates with the amount of co-occurring speciation events, which validates

These results allow phylogeneticist to judge under which circumstances the commonly used speciation model can be safely used.

In a bigger picture, these results showcase the use of a general and flexible method we used to assess the impact of using an oversimplistic tree prior, helping phylogeneticists to find the line between 'too simple' and 'too complex' speciation models.

**Keywords:** computational biology, evolution, phylogenetics, Bayesian analysis, tree prior, pirouette, BEAST2, babette

## 1 Introduction

Modern computational techniques allow to infer phylogenies from genetic data such as DNA, RNA or proteins. BEAST (Drummond & Rambaut 2007) and its descendant BEAST2 (Bouckaert *et al.* 2014) are widely used tools to perform this task, which they can achieve by running a Bayesian analysis given data and tree priors.

BEAST2 gives to the user the option to set up several possible phylogenetic priors (e.g. substitution/clock/speciation models). However, currently available priors can be not suitable to analyze some specific datasets.

For this reason BEAST2 provides users with the possibility to introduce new tree priors, to infer phylogenies based on different assumptions on how the speciation process takes place.

Current phylogenetic tools assume that only a single speciation event can occur at any given time. While this assumption has been proved to be useful

48 to construct a wide variety of successful models (e.g Maddison *et al.* 2007,  
49 Valente *et al.* 2015, Etienne *et al.* 2012, Etienne *et al.* 2014), they disallow  
50 for environmental changes that trigger speciations in multiple clades at a same  
51 point in time.

52 We explore such case introducing the multiple birth model, currently absent  
53 in BEAST. The multiple birth hypothesis aims to include species pump mech-  
54 anisms to investigate drivers and modes of such diversification processes whose  
55 phylogenies show an impressive amount of speciation events in relatively short  
56 times.

57 The (constant-rate) birth-death (BD) model embodies the common assump-  
58 tion that only a single speciation event can occur at any given time. The  
59 multiple-birth-death (MBD) model relaxes this assumption allowing, in ad-  
60 dition to standard BD events, also events in which large-scale environmental  
61 changes lead to speciation bursts. Such hypothesis can be useful to better un-  
62 derstand the history of systems of particular interests for evolutionary biologists,  
63 such as the diversification of cichlid fish in the African Great Lakes (Malawi,  
64 Tanganyika and Victoria), where water level fluctuations are thought to play an  
65 important role in promoting diversification (Verheyen *et al.* 1996, Sturmbauer  
66 *et al.* 2001, Janzen *et al.* 2016, Janzen *et al.* 2017).

67 However, the introduction of new tree priors is not always desirable (Bilder-  
68 beek *et al.* 2019). Current BD tree priors might, in principle, prove to be good  
69 enough at detecting such events despite the lower level of complexity. If this is  
70 the case one should always be more keen to adopt the simplest model.

71 We used the R package `pirouette` (Bilderbeek & Laudanno 2019) to perform  
72 such test, starting on phylogenies simulated under the MBD regime using the  
73 `mbd` package (Laudanno 2018). From such phylogenies we measure the inference  
74 error made adopting a standard BD tree prior in the inference process.

75 With this work we aim, using such inference error distributions, to test  
76 whether or not it is advantageous to implement a new prior model that can  
77 allow the construction of trees where multiple speciations can co-occur at the  
78 same time.

## 79 **2 Methods**

### 80 **2.1 Model**

81 In the MBD model, parameters  $\lambda$  and  $\mu$  correspond, respectively, to the com-  
82 mon per-species speciation and extinction rates present also in the standard BD  
83 model. Additionally, MBD relies on two additional parameters,  $\nu$  and  $q$ .  $\nu$  is  
84 the rate at which an environmental change is triggered. When such event is  
85 triggered, each species present in the phylogeny at that moment has a proba-  
86 bility  $q$  to speciate at that time. This kind of speciation is of a different nature  
87 respect to the one triggered by  $\lambda$ . In fact, whereas parameter  $\lambda$  can be seen  
88 as describing a sympatric process,  $\nu$  induces the rise of geographical barriers  
89 interrupting the gene flow and leading to an allopatric speciation. Even though  
90 multiple speciations can co-occur, polytomies are not allowed in such process as  
91 each species can speciate only once at the time. A likelihood expression for the  
92 process is provided in Laudanno 2018.

### 93 **2.2 Tree simulations**

94 We simulate the speciation process in continuous time using the Doob-Gillespie  
95 algorithm, using the `mbd_sim` function in the `mbd` package (Laudanno 2018). We  
96 let parameters vary using all possible combinations of values as shown in Table 1.  
97 For each parameter setting, we generate 1000 independent reconstructed trees  
98 of the same crown age. We have picked the parameters in such a way that in

the most speciose setting, the simulated trees have usually less than 200 taxa.

Parameter	Values
$\lambda$	(0.2)
$\mu$	(0, 0.15)
$\nu$	(1.0, 1.5, 2.0, 2.5)
$q$	(0.1, 0.15, 0.2)
crown age	6

Table 1: Parameters used to simulate MBD trees. For each parameter setting 1000 trees are simulated.

## 2.3 Inference error estimation

From each MBD tree, we measure the impact of using the simpler BD tree prior, using the `pirouette` R package, as described in detail in Bilderbeek & Laudanno 2019.

In brief: `pirouette` starts from a 'true' (but unobservable in nature) starting phylogeny, from which a DNA sequence alignment (which is observable in nature) is simulated. From each sequence alignment, a Bayesian inference is run, to obtain a posterior distribution of jointly-estimated trees and model parameter estimates. By comparing the true tree and the posterior trees, an inference error distribution is generated. We use the twinning option available in `pirouette` to measure a minimum and full error, in which the minimum error is caused by stochasticity in the full pipeline, where the full error is the added error from using an invalid but standard speciation model.

In our context, the alignments are 1000 nucleotides in length, with a known root sequence of four 250 mono-nucleotide blocks, following the simplest nucleotide substitution model (JC69) and clock model (strict), with a mutation rate of  $\frac{1}{2} \cdot c$ , in which  $c$  is the crown age. With this mutation rate, each nucleotide is expected to mutate (both silently and non-silently) in half of the histories from the root sequence in the past to the sequence in the tips in the present.

119 For the Bayesian inference, we use a JC69 site model, a strict clock model  
 120 and a BD tree prior. Additionally, we use an MRCA prior equal to the crown  
 121 age with a normal distribution of width  $\sigma = 0.01$ . We pick an MCMC setup of  
 122 10M states, which is sampled each 1k moves.

123 For the error measurement, we use the nLTT statistic and a burn-in of 0.1.

124 For the twinning, we let the twin trees follow a BD model.

125 For the experiments, we use both a hand-picked generative model and a  
 126 set of candidate models. We set a BD tree prior, JC69 site model and strict  
 127 clock model as the generative model. We used all other combinations of four  
 128 tree priors, two clock models and five speciation models, resulting in a set of 39  
 129 candidate models.

### 130 **3 Results**

131 The inference error made for each of the parameter combinations is shown in  
 132 Fig. 1. For both extinction rates, we find, as expected, that the error increases  
 133 for increased  $\nu$  or  $q$ . Also in line with our predictions, we find no difference  
 134 between the two extinction rates.

### 135 **4 Discussion**

136 From the four MBD parameters  $\lambda$ ,  $\mu$ ,  $\nu$  and  $q$ , we investigated 1, 2, 4 and 3  
 137 different values respectively. We chose to use only one  $\lambda$ , as the proportion  
 138 of species created in a co-occurrent speciation event is dependent on the ratio  
 139 between  $\lambda$  and a combination of  $\nu$  and  $q$ .

140 We selected our parameters in such a way that the simulated trees had  
 141 usually less than 200 taxa. One could argue that starting from trees with more  
 142 taxa would result in a clearer inference, which we agree upon. We chose to

## Inference errors in parameter space

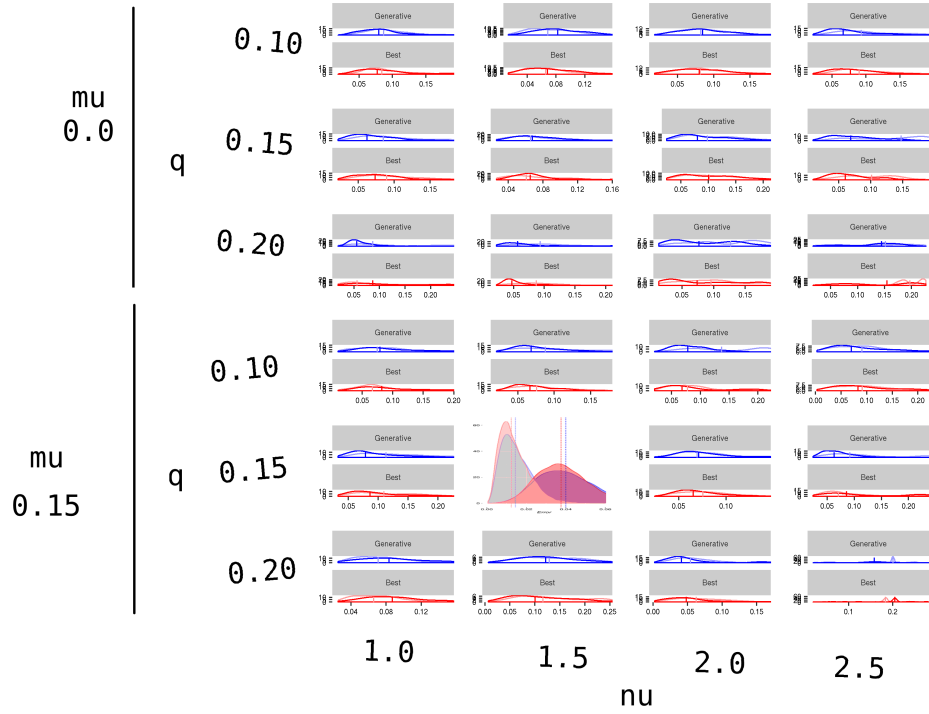


Figure 1: The inference error distribution for the different biological parameter settings. In all cases,  $\lambda = 0.2$  and crown age equals 6.

143 use more replicates over more taxa, as we could easily add more replicates in a  
144 scheduled way.

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