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Local and relaxed clocks, the best of both worlds

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Time-resolved phylogenetic methods use information about the time of sample collection to estimate the rate of evolution. Originally, the models used to estimate evolutionary rates were quite simple, assuming that all lineages evolve at the same rate, an assumption commonly known as the molecular clock. Richer and more complex models have since been introduced to capture the phenomenon of substitution rate variation among lineages. Two well known model extensions are the local clock, wherein all lineages in a clade share a common substitution rate, and the uncorrelated relaxed clock, wherein the substitution rate on each lineage is independent from other lineages while being constrained to fit some parametric distribution. We introduce a further model extension, called the flexible local clock (FLC), which provides a flexible framework to combine relaxed clock models with local clock models. We evaluate the flexible local clock on simulated and real datasets and show that it provides substantially improved fit to an influenza dataset. An implementation of the model is available for download from <https://www.github.com/4ment/flc>.

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ABSTRACT

Time-resolved phylogenetic methods use information about the time of sample collection to estimate the rate of evolution. Originally, the models used to estimate evolutionary rates were quite simple, assuming that all lineages evolve at the same rate, an assumption commonly known as the molecular clock. Richer and more complex models have since been introduced to capture the phenomenon of substitution rate variation among lineages. Two well known model extensions are the local clock, wherein all lineages in a clade share a common substitution rate, and the uncorrelated relaxed clock, wherein the substitution rate on each lineage is independent from other lineages while being constrained to fit some parametric distribution. We introduce a further model extension, called the flexible local clock (FLC), which provides a flexible framework to combine relaxed clock models with local clock models. We evaluate the flexible local clock on simulated and real datasets and show that it provides substantially improved fit to an influenza dataset. An implementation of the model is available for download from <https://www.github.com/4ment/flc>.

INTRODUCTION

Phylogenetic methods provide a powerful framework for reconstructing the evolutionary history of viruses, bacteria, and other organisms. Correctly estimating the rate at which mutations accumulate in a lineage is essential for phylogenetic analysis, as the accuracy of inferred rates can heavily impact other aspects of the analysis. Classic approaches to infer the substitution rate of a group of organisms rely on the existence of a so-called "molecular clock". The molecular clock hypothesis dictates that mutations accumulate at an approximately steady rate over time, implying that the genetic distance between two organisms is proportional to the time since these organisms last shared a common ancestor. The molecular clock hypothesis was first proposed almost 50 years ago by Emile Zuckerkandl and Linus Pauling (Zuckerkandl and Pauling, 1965) who suggested that the substitution rate was effectively constant over time. This very restricted model of evolution has been implemented using a "strict clock" model in phylogenetic inference software, but the rates of evolution in many organisms appears to change over time and the model can not capture this phenomenon.

In recent years, richer models have been developed to capture the complexity of the evolutionary process. Sanderson (2002) and Thorne et al. (1998) proposed to model rate heterogeneity among lineages using auto-correlated clock models using penalized likelihood and Bayesian inference respectively. In these parameter rich models, the rate of each lineage is assumed to be correlated to that of the parent lineage. The auto-correlation assumption could be justified by considering that the substitution rate is influenced by heritable mechanisms such as metabolic rate or generation time. However there is no guarantee that rates evolve in an auto-correlated manner, especially when the timescale under study is relatively small Drummond et al. (2006). An alternative approach is to assume that substitution rates on adjacent branches are independent draws from an underlying parametric distribution. Drummond et al. (2006) chose to forgo the hierarchical Bayesian framework and opted for a likelihood approach that requires the rates to fit a discretized distribution. The log-normal and exponential distributions are commonly used as they are available in the widely used BEAST package (Drummond and Rambaut, 2007; Bouckaert et al., 2014). The auto-correlated and uncorrelated clock models are referred to as *relaxed*

47 *clock models* due their ability to relax the constant rate assumption.

48 Local clock models are an alternative to relaxed clocks, where the model assumes that the substitution
49 rate is constant within a clade but can differ between clades (Yoder and Yang, 2000; Yang and Yoder,
50 2003). The number and locations of these local clocks can be inferred from the data using the random
51 local clock model (Drummond and Suchard, 2010) or local clocks can be assigned by the user based on
52 prior information.

53 In this manuscript we introduce a hybrid model that integrates features of both the local and the
54 relaxed clock models. In the model each local clock can be specified either as a strict clock (as in the
55 original formulation of the local clock model) or as a relaxed clock. We call this model the flexible local
56 clock (FLC) model. We evaluate the FLC model using a newly implemented module for the BEAST2
57 package, which uses Markov chain Monte Carlo to carry out inference of model parameters (Bouckaert
58 et al., 2014). We reanalyzed an influenza virus (Drummond and Suchard, 2010) and a HIV (Wertheim
59 et al., 2012) data set to evaluate the utility of the FLC model and compared its fit to the data to that given
60 by other models.

61 METHODS

62 Phylogenetic packages such as BEAST provide several options to model lineage-specific rate variation,
63 known as heterotachy, without overfitting the model. One of the first ingredients of the FLC model is the
64 uncorrelated relaxed clock model Drummond et al. (2006), arguably the most popular lineage-specific
65 rate model. The uncorrelated relaxed clock model uses a single discretized parametric distribution to
66 model rate heterogeneity. In the original formulation of the model, a parametric distribution, usually
67 lognormal, is discretized into a fixed number of components, with the number of these components equal
68 to the number of branches b in the tree. In its simplest form, the model assumes a one-to-one relationship
69 between a rate at a branch and one of the components. For a lognormal distribution, this approach only
70 requires estimating two parameters (i.e. mean and standard deviation) instead of $2N - 2$ parameters if a
71 hierarchical model was used, where N is the number of sequences.

72 The other ingredient of the FLC model is the local clock which was first proposed by Yoder and Yang
73 (2000). This model allows lineages within a region of the tree to evolve at exactly the same rate. We
74 define a local clock on a phylogeny as a monophyletic group where the substitution rate of every lineage
75 is equal. As in Drummond and Suchard (2010), we assume the existence of another clock (e.g. a 'global'
76 clock) for lineages that are not assigned a local clock.

77 Herein, we propose to relax the constraint that lineages within a local clock evolve at exactly the same
78 rate by replacing this implicit strict clock by a relaxed clock.

79 We applied the FLC model to two data sets of heterochronous viral nucleotide sequences. The first data
80 set comprises an alignment of 69 human influenza A/H3N2 virus haemagglutinin (HA) sequences (987 nt
81 in length) isolated between 1981 and 1998. The evolutionary rates and time to the most recent ancestors
82 (tMRCAs) of this data set was previously investigated using a random local clock method (Drummond
83 and Suchard, 2010) with a Bayesian Markov chain Monte Carlo (MCMC) approach implemented in
84 BEAST (Drummond and Rambaut, 2007). We reanalysed the data using BEAST with either the FLC
85 model, uncorrelated lognormal relaxed clock (UCLN), local clock (LC), and a random local clock model
86 (RLC). As in the original study, our analyses use the HKY+ Γ_4 substitution model that incorporates
87 gamma-distributed rate variation among sites (4 rate classes). The FLC and LC models require manual
88 assignment of each lineage to a local clock with the appropriate constraints on the phylogeny. Drummond
89 and Rambaut (2007) noticed that the substitution rate of the lineages comprising viruses sampled after
90 1990 appeared higher than the pre-1990 lineages. We therefore assigned sequences sampled after 1990 to a
91 local clock for both LC and FLC models. For each local-based model, we conducted two separate analyses
92 in which the branch subtending the clade containing the late viruses (1990-onward) were assigned either
93 to a local clock or the ancestral rate. We specified a diffuse prior on the substitution rates of the LC and
94 FLC models using an exponential distribution with a mean of 0.003. For the log-normal distribution
95 of the relaxed clock we used an exponential prior distribution ($\lambda = 1/0.003$) on the mean parameter
96 and an exponential prior distribution ($\lambda = 1/0.33$) on the standard deviation parameter. As in the study
97 describing the RLC model we used a Poisson distribution with $\lambda = \log 2$ as a prior on the number of local
98 clocks, thereby placing 50% prior probability on a single rate across the phylogeny. Finally, we assumed *a*
99 *priori* that rate multipliers are independently gamma distributed with $\alpha = 0.5$ and $\beta = 2$ as in Drummond
100 and Suchard (2010).

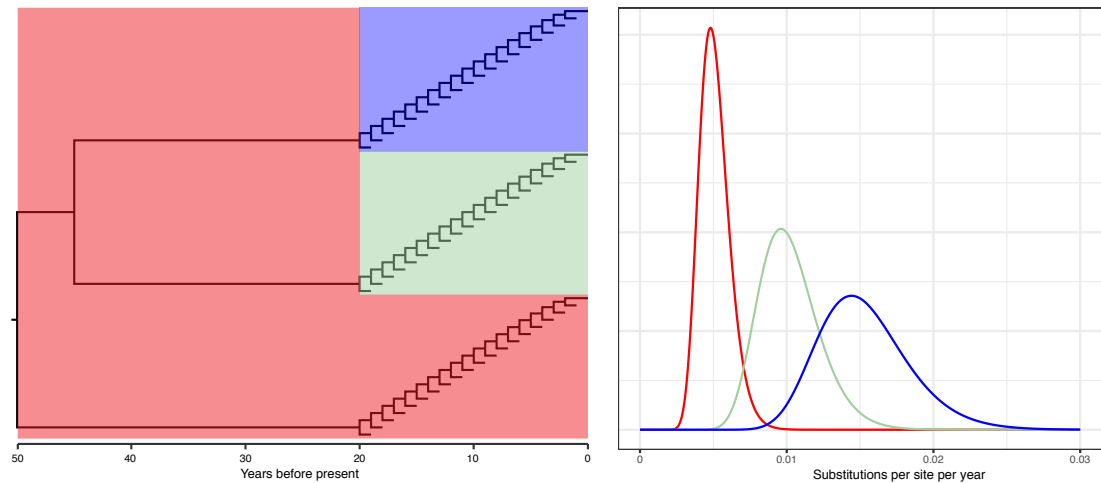


Figure 1. Phylogenetic tree and substitution rates used to simulate data sets.

101 For each data set, we calculated the marginal likelihood of the data under each model using the
 102 stepping stone algorithm to compare competing models Xie et al. (2011). We used a series of 100 power
 103 posteriors where β values are chosen to be evenly spaced quantiles of a Beta distribution with parameters
 104 $\alpha = 0.3$ and $\beta = 1.0$. These parameters results in half of the power posteriors being evaluated for $\beta < 0.1$
 105 for which the power posterior is changing rapidly, as suggested by Xie et al. (2011). Each MCMC was
 106 run for 10 million iterations and the first 10% of the samples were discarded as burn-in.

107 Simulations

108 To validate the implementation of our model we simulated data sets using the FLC model. Our approach
 109 is similar to a simulation-based study (Worobey et al., 2014) that showed that the local clock model is best
 110 suited to model rate variation among influenza virus sequences sampled from three different hosts (i.e.
 111 equine, human, and birds). Worobey *et al.* assigned different local clocks to each of the monophyletic
 112 bird and human clades and simulated nucleotide alignments containing 10,000 sites. Phylogenies were
 113 estimated using either a strict, flexible local or local clock model using the BEAST package (Drummond
 114 and Rambaut, 2007). The simulations showed that only the local clock model was able to recover the true
 115 tree. In this study, we used the same topology and divergence times, and replaced standard local clocks
 116 with flexible local clocks.

117 10 replicates containing 10,000 sites were simulated using the program simultron (Fourment and
 118 Holmes, 2014) under the HKY model ($\kappa = 3$ and equal nucleotide frequencies). The standard deviation
 119 σ of the lognormal distributions were all set to be equal to 0.2. The μ parameter of the lognormal
 120 distributions of the equine, human, and bird clades were set such as the mean of the distributions were
 121 5×10^{-3} , 1×10^{-2} , and 1.5×10^{-2} respectively (Figure 1). The choice of the parameters results in
 122 roughly bell shaped distributions centered on the substitution rates used in the Worobey *et al.* study. We
 123 analyzed the simulated data sets with the HKY model and the skyline coalescent tree prior under the
 124 strict, flexible local, local, relaxed, and random local clocks. The simulation script is available from
 125 <http://www.github.com/4ment/flc-data>.

126 RESULTS AND DISCUSSION

127 We analyzed the influenza virus data set with BEAST under a variety of models including the FLC model.
 128 Since the flexible local clock can be composed of a combination of strict and relaxed clocks, we specify
 129 the type of clock between brackets. For example, we use FLC [strict&UCLN] to denote a flexible local
 130 clock with a strict clock on the early lineages (i.e. sequences before 1990) and a uncorrelated lognormal
 131 relaxed clock (UCLN) on the later lineages. For local and flexible local clocks we can specify whether
 132 the branch leading to the clade with a local clock should be included in the new clock (contains the stem).
 133 To test which models better fit to the data we calculated the marginal likelihood for each model (Table 1).

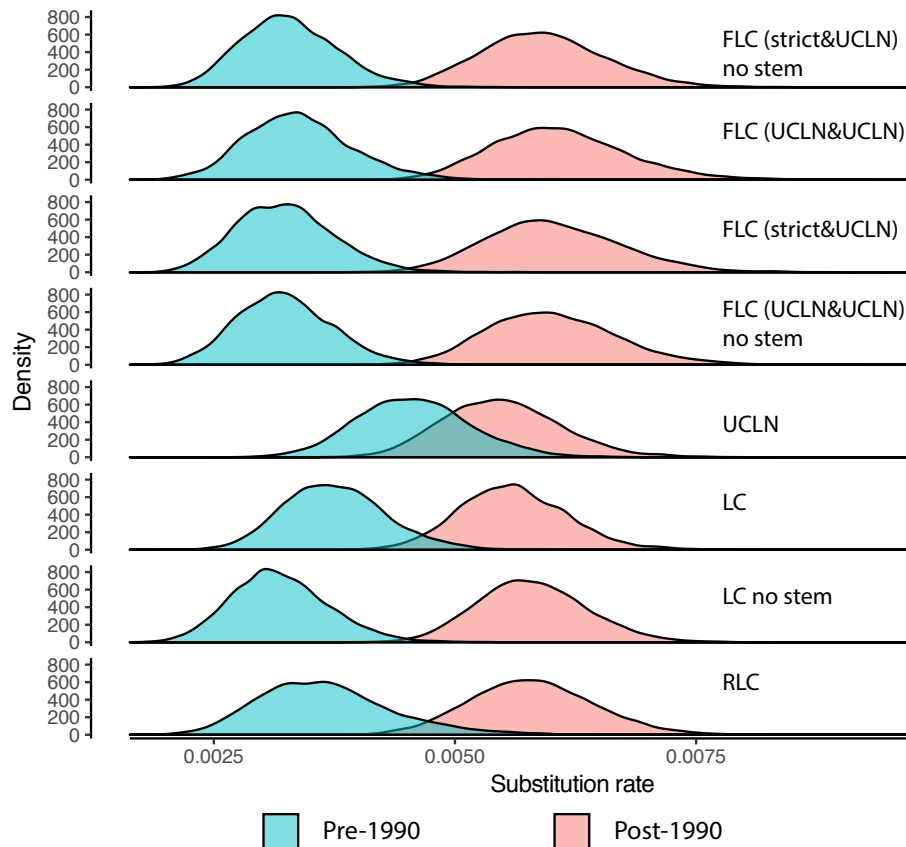


Figure 2. Posterior distributions of the mean substitution rate of the lineages comprising viruses sampled after 1990 and the pre-1990 lineages. UCLN: uncorrelated lognormal relaxed clock, RLC: random local clock, FLC: flexible local clock LC: local clock. For the local clock models labeled no stem, the branch subtending the post 1990 clade is not assigned to the local clock.

Model	Marginal likelihood	Contains stem
FLC [strict&UCLN]	-4381.72	no
FLC [UCLN&UCLN]	-4382.07	yes
FLC [strict&UCLN]	-4382.92	yes
FLC [UCLN&UCLN]	-4383.28	no
UCLN	-4385.04	NA
LC	-4386.81	yes
LC	-4387.69	no
RLC	-4415.36	NA

Table 1. Marginal likelihoods calculated using the stepping stone algorithm. UCLN: uncorrelated lognormal relaxed clock, RLC: random local clock, FLC: flexible local clock LC: local clock. The "Contains stem" column specifies whether the branch subtending the post-1990 clade is assigned to the local clock.

134 As in the original study (Drummond and Suchard, 2010), every model shows a substitution rate
135 increase in sequences sampled after the 1990 (Figure 2).

136 The marginal likelihood estimates (Table 1) suggest that the best models are the FLC models, followed
137 by the UCLN, LC, and RLC models. The inclusion of the stem in the FLC and LC models appears to a
138 minor effect on the model fit depending on the model, but the marginal likelihood estimates are subject to
139 Monte Carlo error and caution should be exercised in order to avoid overinterpreting small differences.
140 The 95% highest posterior density (HPD) of the standard deviation of the lognormal distribution assigned
141 to the global clock includes zero, suggesting that there is little rate variation outside the post-1990 clade
142 (i.e. FLC [UCLN&UCLN]). It is therefore no surprise that the marginal likelihoods of the FLC models
143 with a strict or UCLN clock on the pre-1990 lineages are similar. Interestingly the UCLN model appears
144 to fit better to the data than the RLC and LC clocks.

145 Results on simulated data

146 We simulated 10 data sets under the flexible local model and estimated the phylogenies using several
147 clock models. The comparison of the maximum clade credibility (MCC) tree to the true topology reveals
148 that the strict and relaxed clock models could not recover the rooting of the true tree in any replicate.
149 Interestingly the 95% HPD intervals of the root node age contained the true value in four and two of the
150 replicates using the relaxed and strict clock models, respectively. The MCC trees of the standard local
151 clock model recovered the true rooting and the root age was recovered in the 95% HPD in only three
152 replicates. The MCC trees of the flexible local clock model had the same rooting as the true tree in 9 out
153 of 10 cases and the 95% HPD of the root age contained the true value for 8 out of 10 replicates.

154 Limitations and conclusions

155 As in the standard local model, the flexible local clock model introduced in this paper assumes that
156 the user knows the number and the location of the rate shifts. Drummond and Suchard (2010) devised
157 the random local clock to address this limitation using a stochastic search variable selection method
158 to sample over random local clocks. Unfortunately that approach is not easily amenable to integration
159 with the FLC model since the substitution rate within a clock can either be constant or heterogeneous
160 across lineages. Although it should be possible to use reversible jump MCMC to sample the posterior
161 distribution it is not clear how to deal with a variable number of lineages assigned to a relaxed clock.
162 For example, the assignment of a relaxed clock with a two parameter distribution to a single branch
163 would over-parametrize the model. An interesting direction for further research would be to develop an
164 algorithm that automatically selects the clock type for each local clock.

165 The FLC model is implemented in the BEAST 2 package as a plugin and is available from
166 <https://www.github.com/4ment/flc>.

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