

Bayesian Phylogenetic Inference using RevBayes:

Model Adequacy Testing

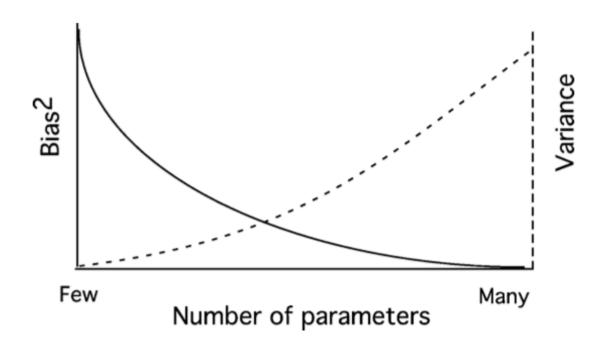
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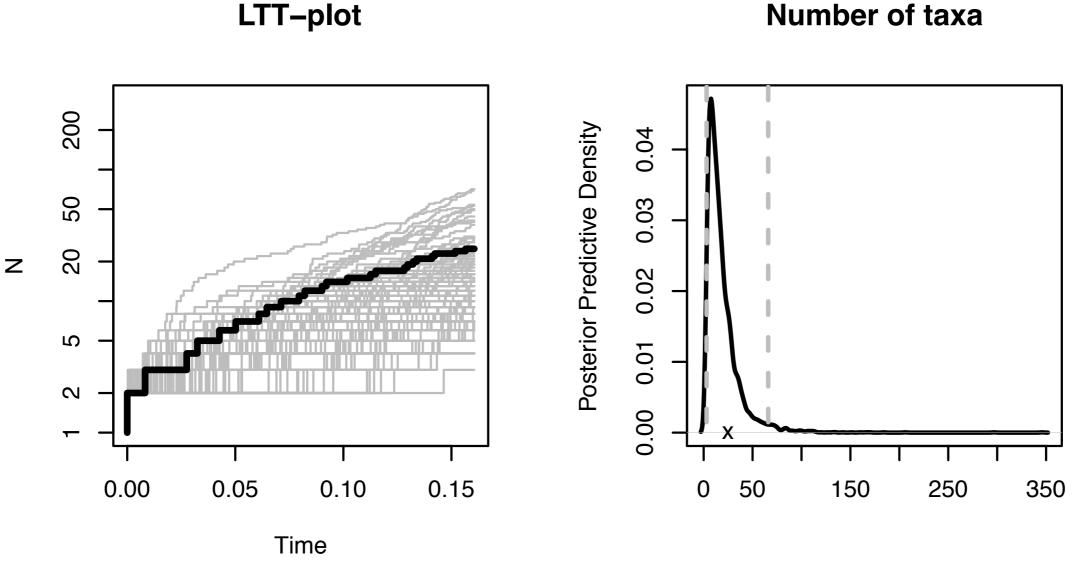
Model Specification Issues

Model selection, adequacy, and related issues

The model is central to parametric estimation of phylogeny: an under parameterized model will cause estimates to be biased (e.g., under estimation of branch lengths, topological error, inflated estimates of nodal support...); however, an over parameterized model will inflate estimation error (error variance, etc.).

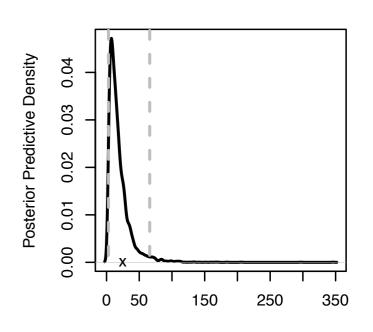


What is the probability of the data under the model? Is the data an outlier and improbable under this model?



Consider that our data are the number of species and the model is a birth-death process for species diversification. This example shows that the data is not unexpected under the model.

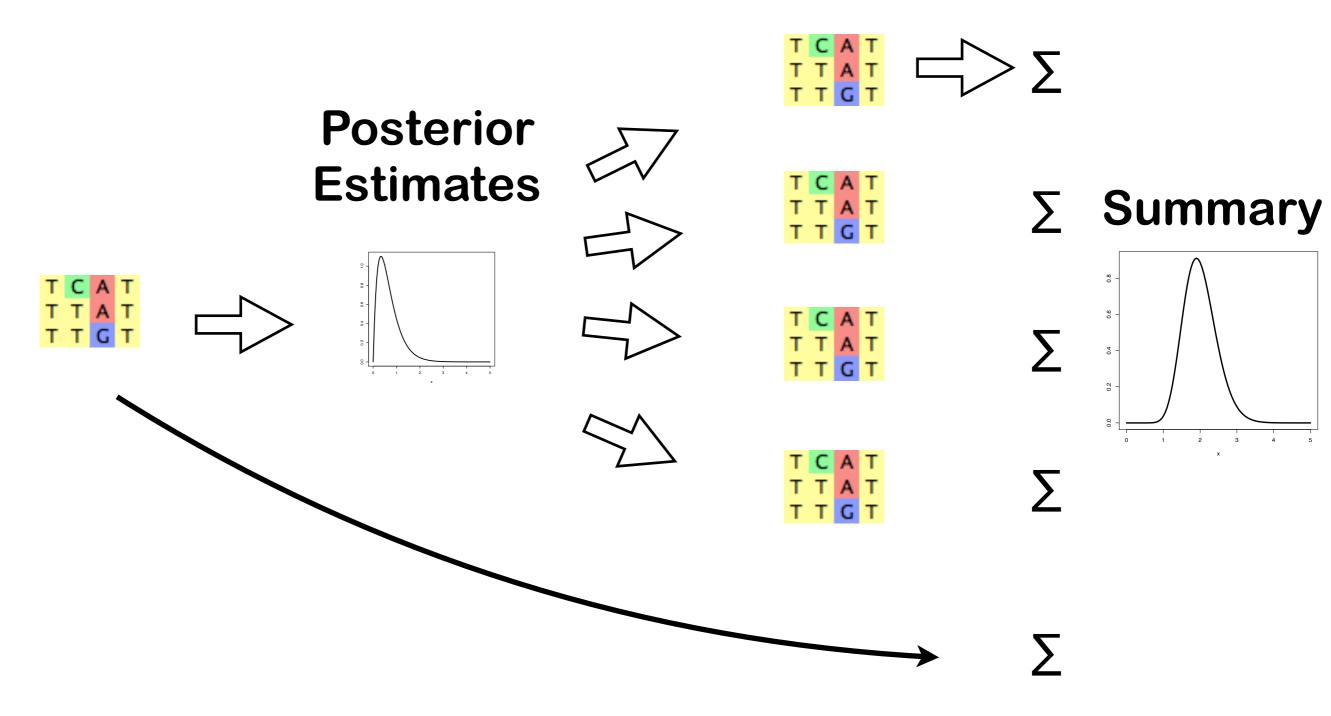
In simple cases, like the number of species under a birth-death model, we can compute the distribution of the data analytically.

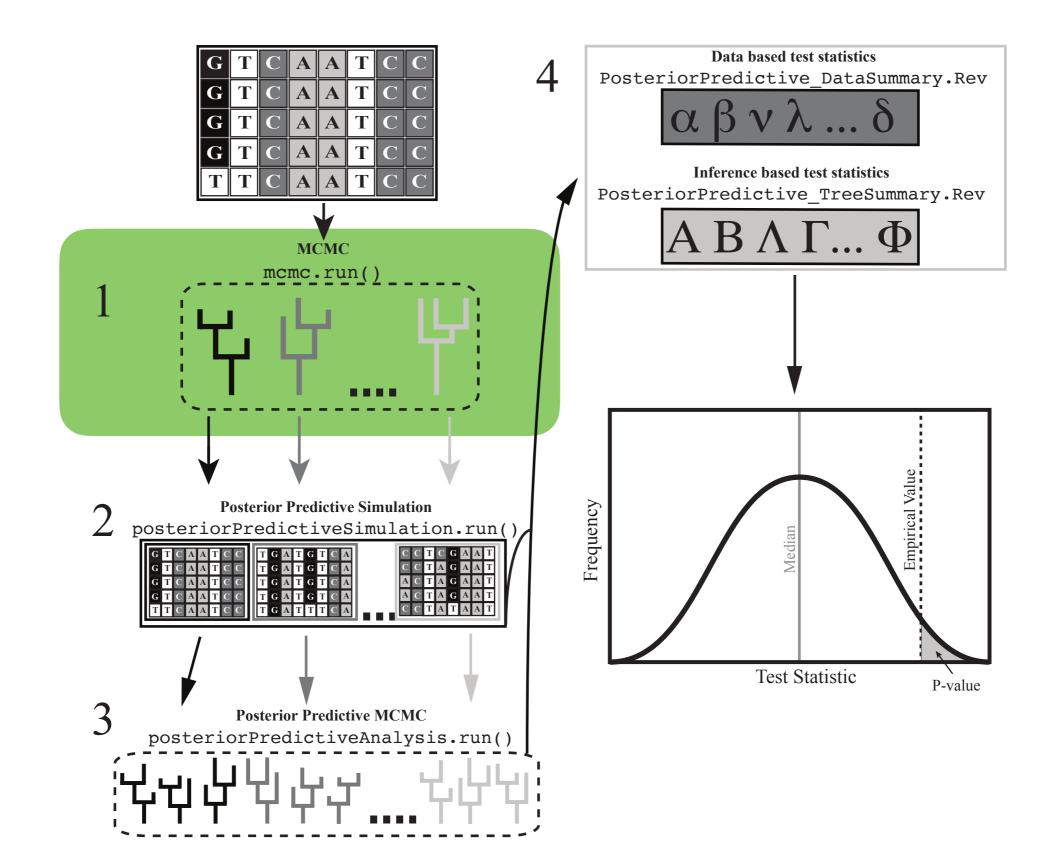


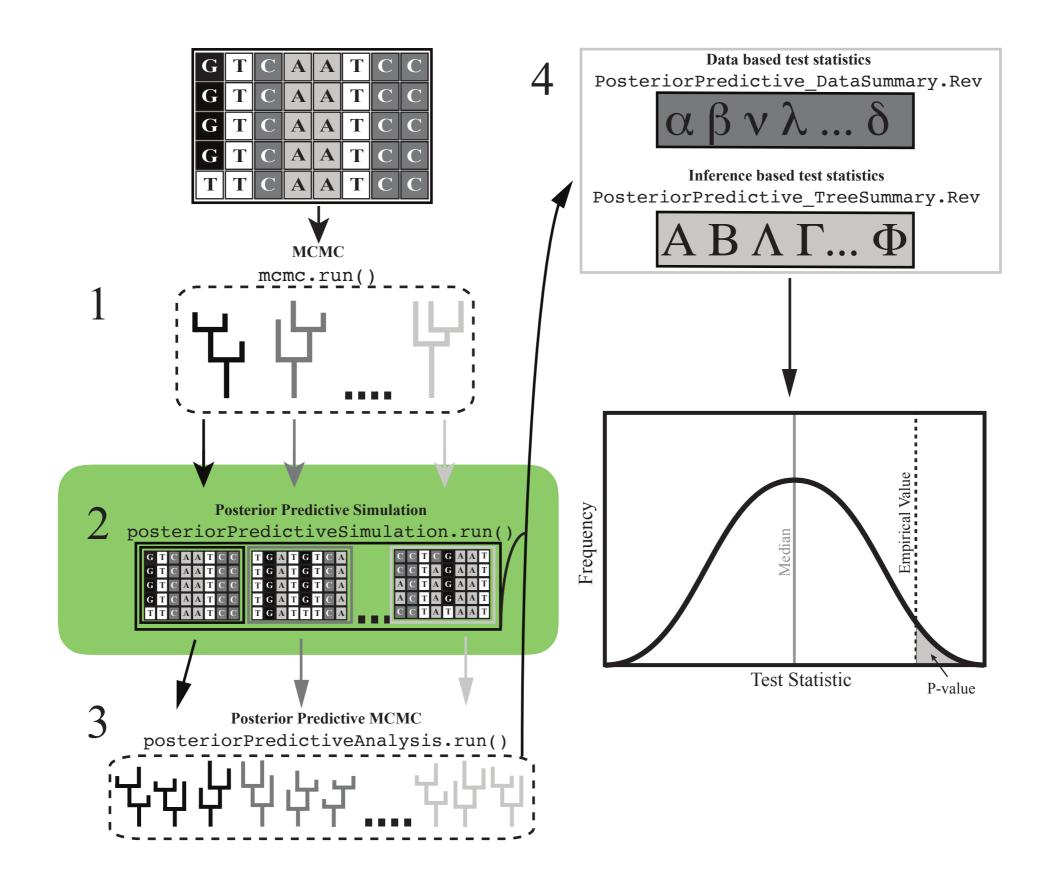
But what should we do in complex situations, like multiple sequence alignments, to compare the data?

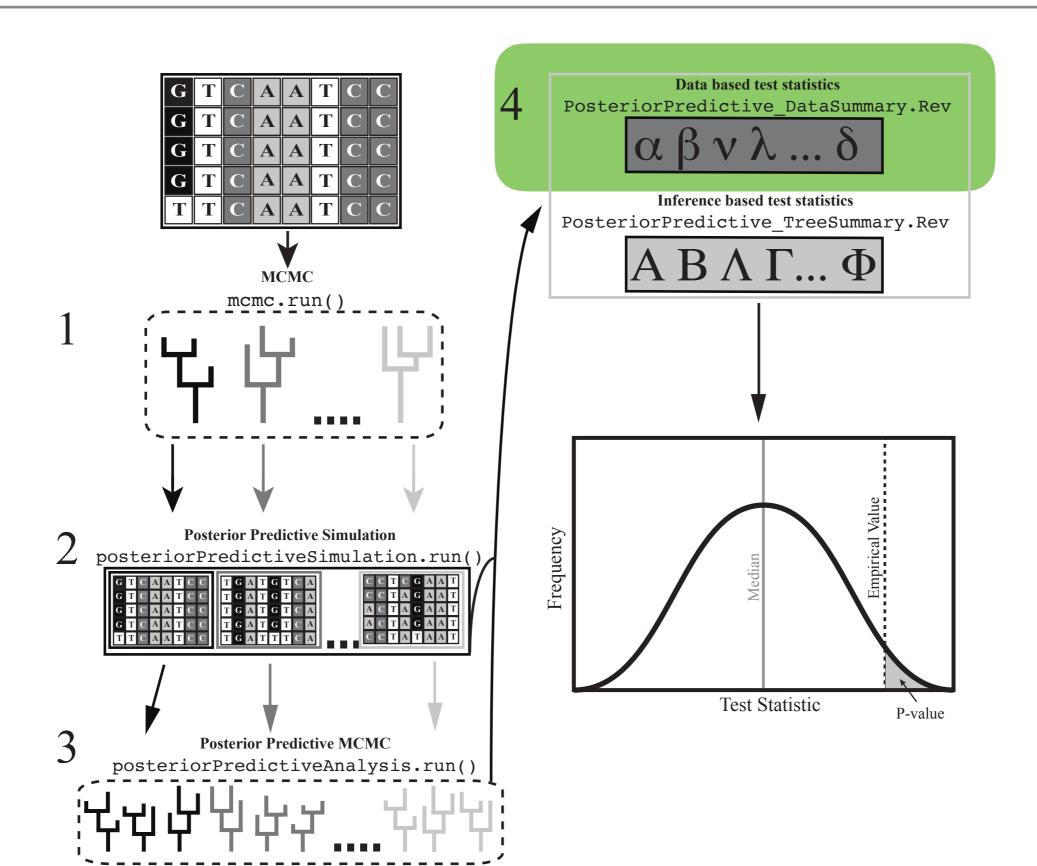
We use simulations and summary statistics!

Simulation









Example data-based summary statistics

Number of invariant sites:

Tests the rate of evolutions.

Maximum, Minimum & Mean GC content:

Tests for base composition and outliers in base composition

Variance of GC content:

Variation on base composition and heterogeneity of substitution process

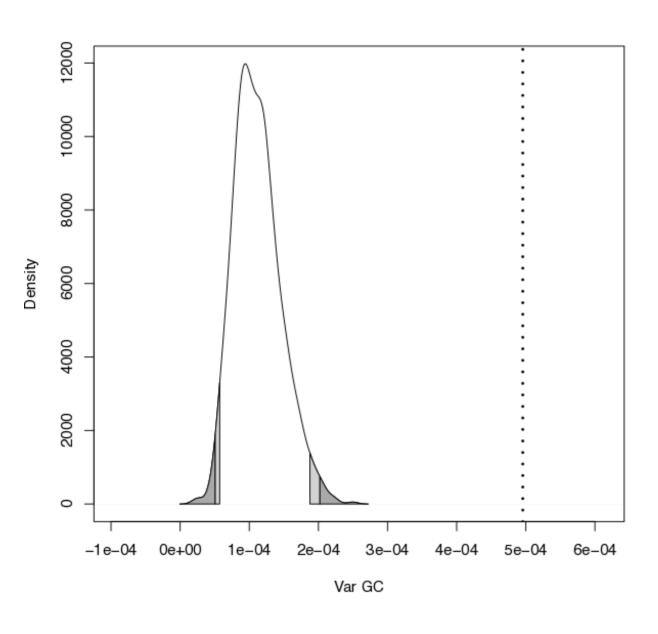
Maximum pairwise distance:

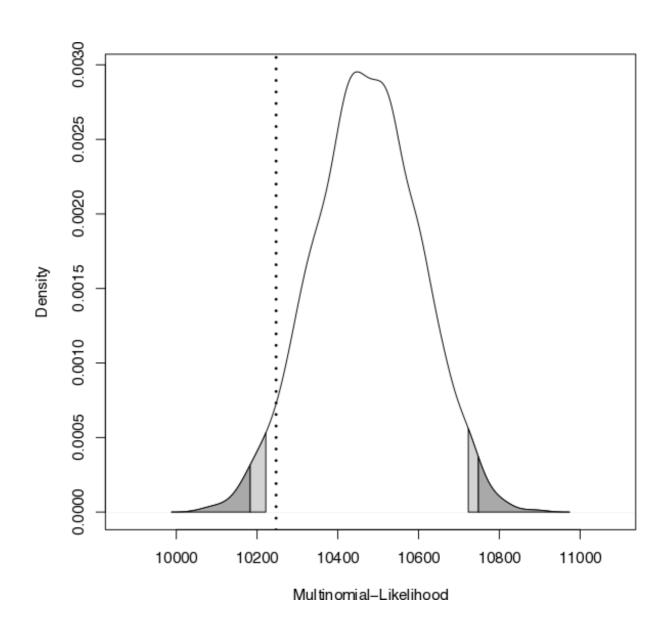
Tests for the rate of evolution along branches.

Multinomial-likelihood:

General test of model fit.

Example Analysis





Computing P-Values

Lower one-tailed p-value:

$$p_L = P(s(d) \le s(e)) = \frac{1}{M} \sum_{i=1}^{M} \mathbf{1}_{s(d_i) \le s(e)}$$

Upper one-tailed p-value:

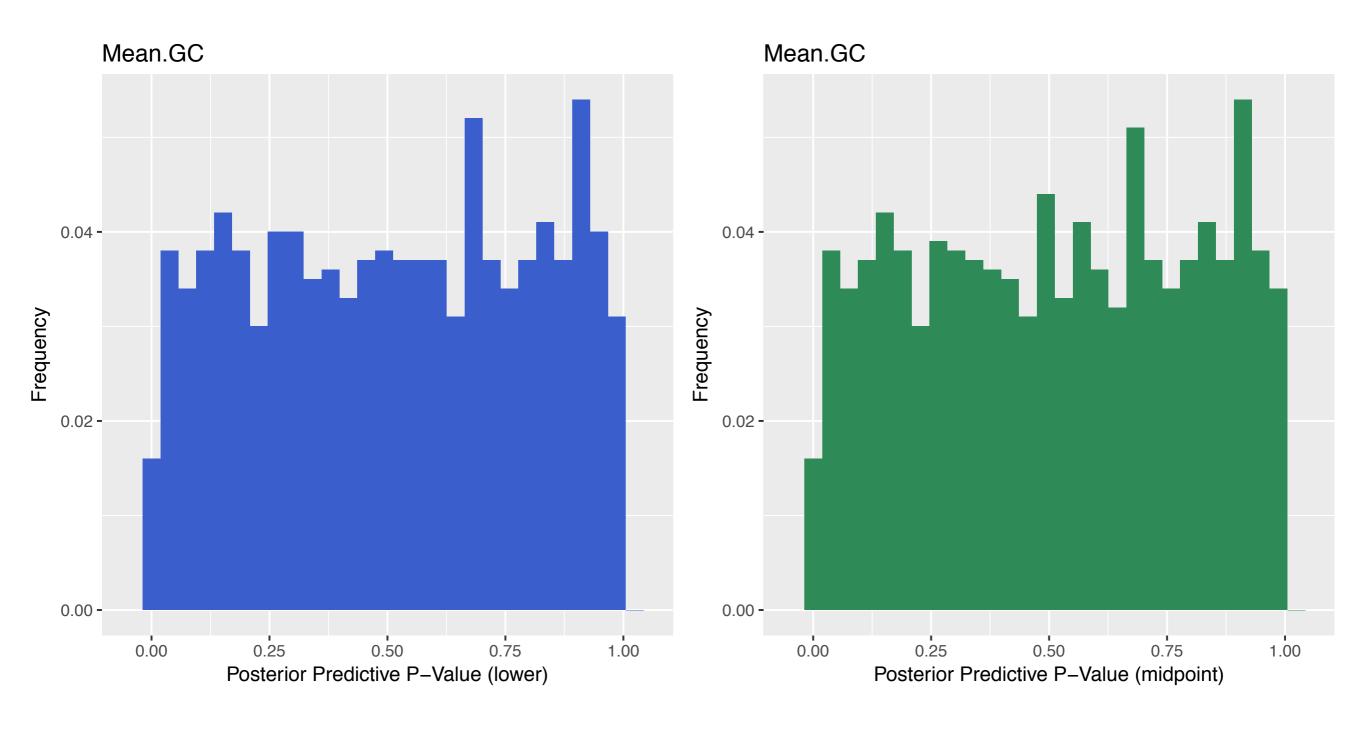
$$p_U = P(s(d) \ge s(e)) = \frac{1}{M} \sum_{i=1}^{M} \mathbf{1}_{s(d_i) \ge s(e)}$$

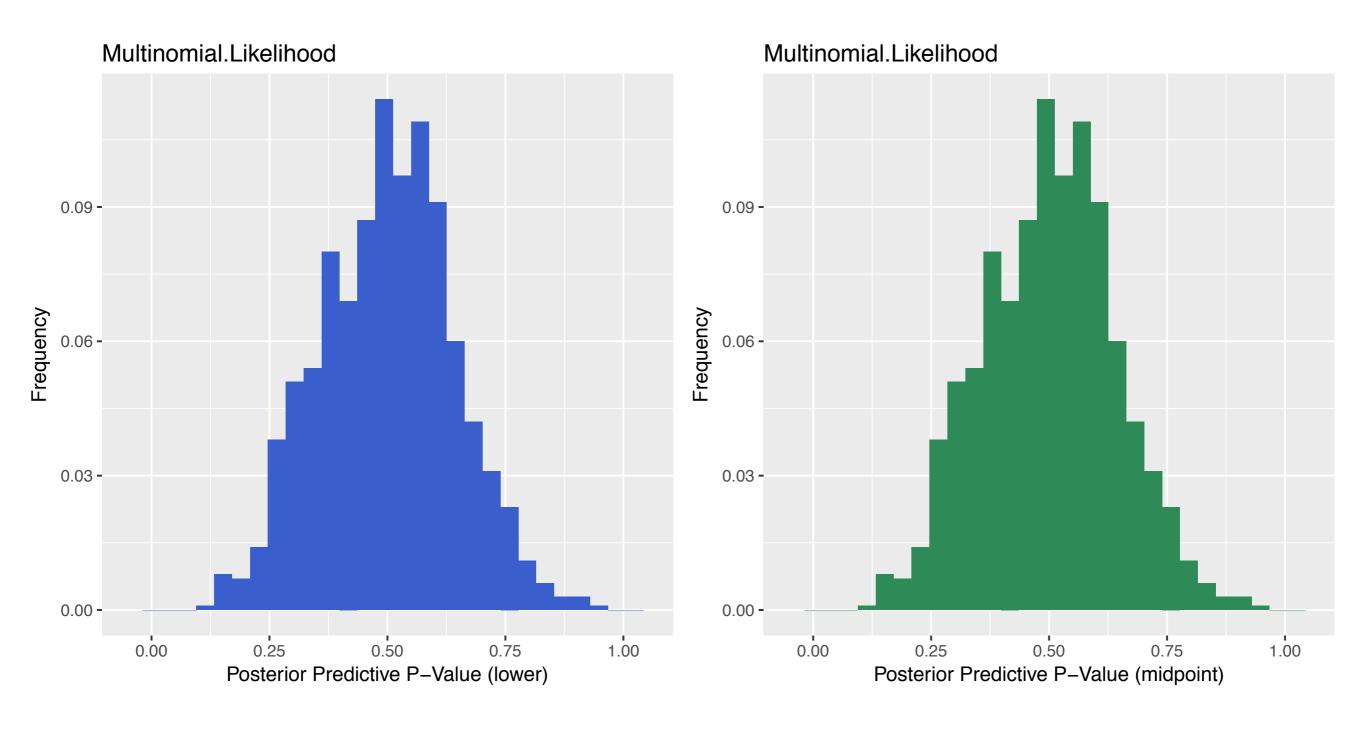
Two-tailed p-value:

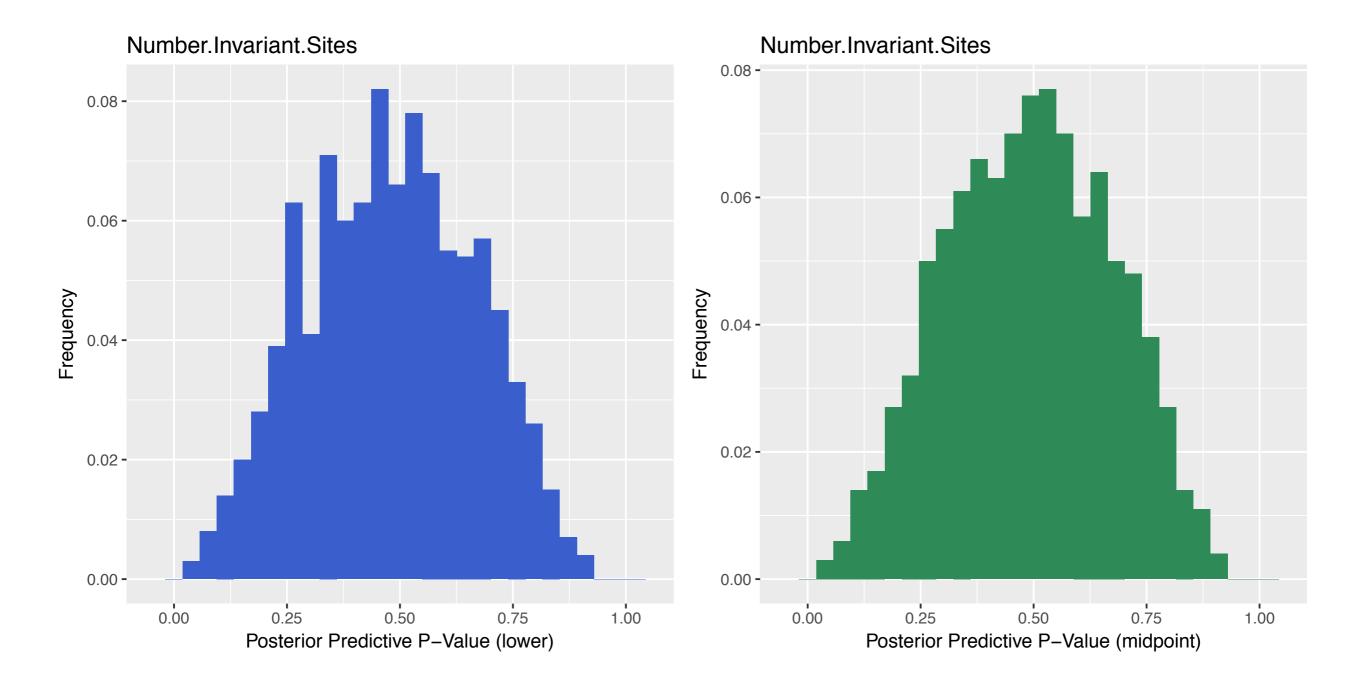
$$p_T = 2 \times \min(p_L, p_U)$$

Midpoint lower one-tailed p-value:

$$p_M = P((d) \lesssim s(e)) = \frac{1}{M} \sum_{i=1}^{M} \mathbf{1}_{s(d_i) < s(e)} + \frac{1}{2M} \sum_{i=1}^{M} \mathbf{1}_{s(d_i) = s(e)}$$







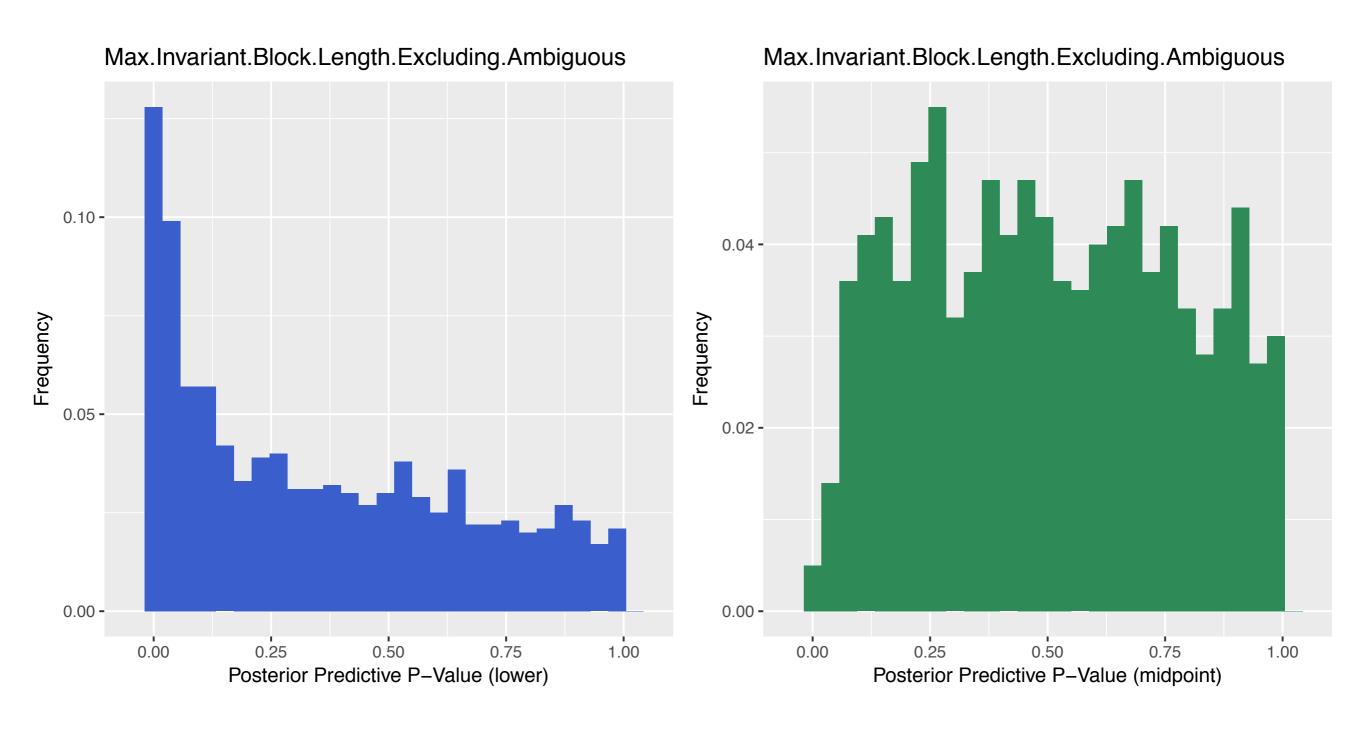
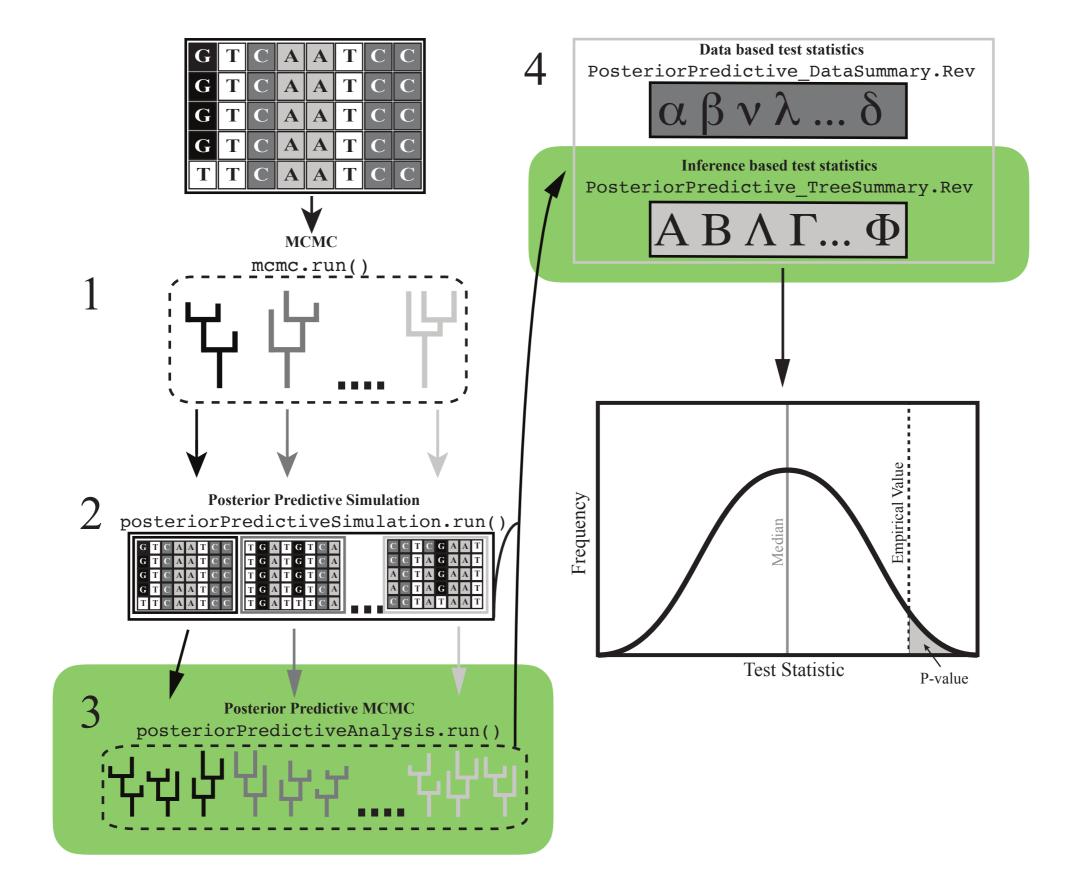
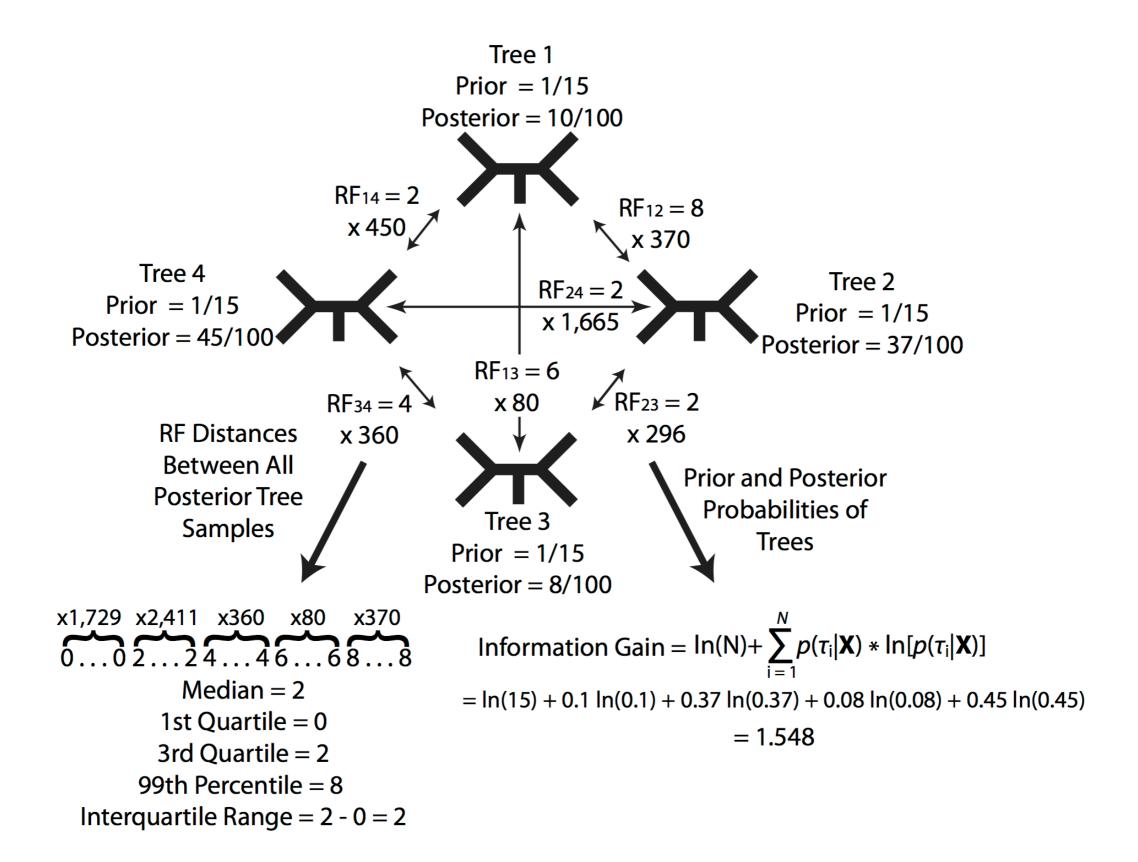


Table 2: Lower 1-tailed midpoint p-values for data based test statistics.

Statistic\Model	JC	GTR	GTR + Inv	GTR + Gamma	GTR + Gamma + Inv
#Invariant Sites	1	1	0.903	0.8535	0.224
Max(GC)	0	1	0.9815	0.683	0.9355
Max(PD)	0	0	0	0.0885	0.3075
Min(GC)	0	0.984	0.5455	0.088	0.341
Min(PD)	0.3685	0.8665	0.554	0.7205	0.544
Mean(GC)	0	0.999	0.7485	0.158	0.5385
Var(GC)	1	1	1	1	1
Multinomial- Likelihood	1	1	0.879	0.535	0.043





Example inference-based summary statistics

Mean and Quantiles Robinson-Foulds-Distance

Captures the spread of the posterior distribution on trees.

Mean and variance tree-length:

Tests for inferred branch lengths and the expected number of substitutions

Entropy:

comparing the prior to the posterior distribution of phylogenetic tree topologies

Table 1: Lower 1-tailed midpoint p-values for inference based test statistics.

Statistic\Mo del	JC	GTR	GTR + Inv	GTR + Gamma	GTR + Gamma + Inv
mean(RF)	1	0.974	0.981	0.931	0.772
q(RF,0.25)	0.5	0.499	0.9335	0.755	0.4905
q(RF,0.5)	0.999	0.4795	0.9625	0.829	0.5795
q(RF,0.75)	1	0.868	0.9835	0.8885	0.6685
q(RF,0.99)	1	0.9995	0.995	0.9925	0.755
q(RF.0.999)	1	1	0.9985	0.969	0.823
mean(TL)	0.426	0.3855	0.51	0.945	0.887
var(TL)	0.391	0.998	0.911	1	1
Entropy	0.999	1	0.99	0.992	0.972

