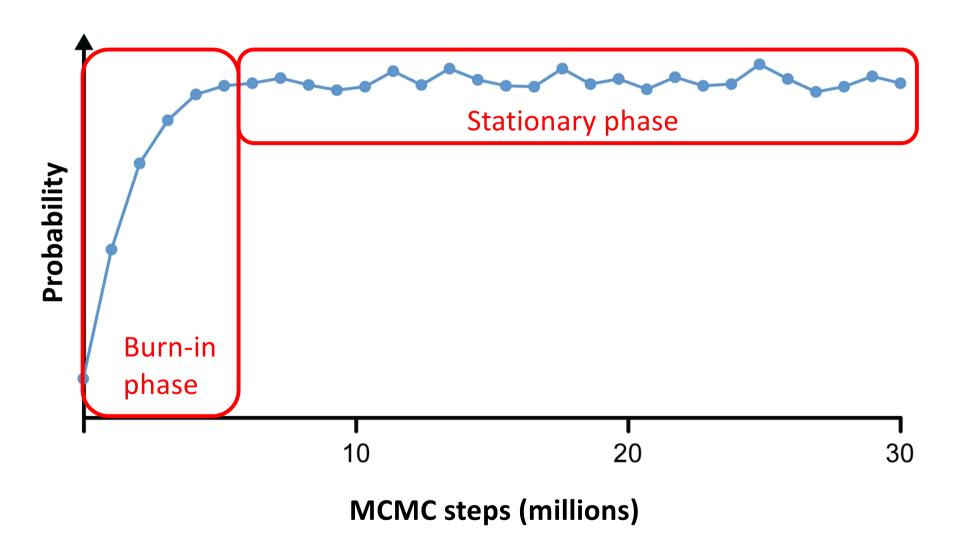
Lecture 2.3

Bayesian Phylogenetics II

MCMC Diagnostics

Samples from the MCMC



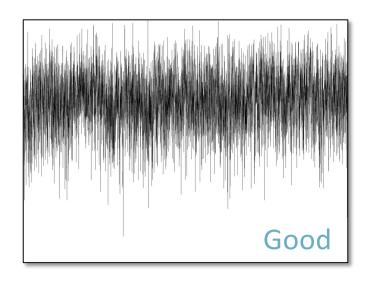
Diagnostics

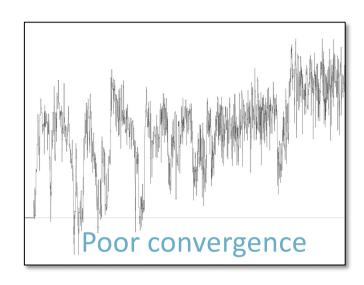
1. Convergence

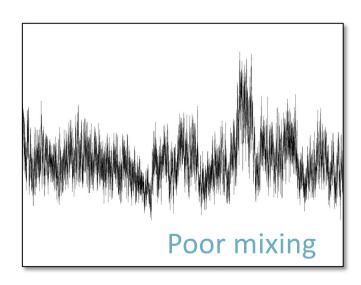
Are we drawing samples from the stationary distribution?

2. Sufficient sampling

Have we drawn enough samples to allow a reliable estimate of the posterior distribution?







Convergence

- Run at least 2 (preferably more) independent chains
- Likelihoods should be similar
- Estimates of model parameters should be similar

Sufficient sampling

- Effective sample size (ESS)
 Have we drawn enough independent samples to produce a reliable estimate of the posterior distribution?
- ESS is preferably >200 for each parameter
- ESS can be increased by:
 - Increasing the length of the MCMC
 (and decreasing the frequency of sampling accordingly)
 - Modifying the MCMC proposals

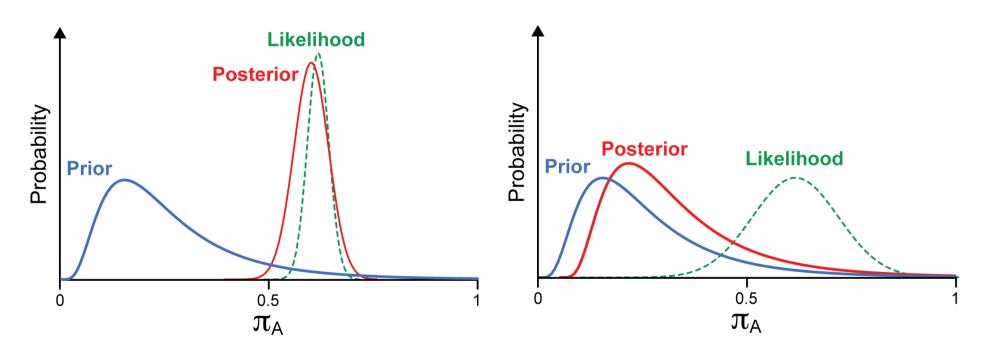
Advantages and Problems

Advantages

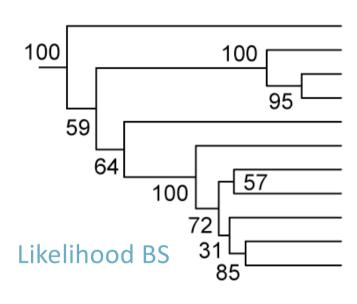
- Able to implement complex (highly parameterised) models
- Posterior probabilities have an intuitive interpretation
- Can incorporate independent information (in the prior)
- Integrate over 'nuisance' parameters

Influence of priors

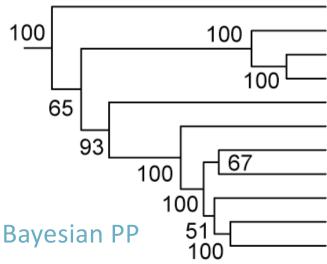
- Sensitivity of the posterior to the prior
- This problem can occur if the data are uninformative, the prior is strong, or both



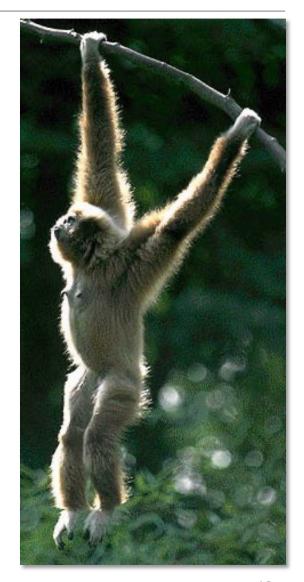
Node support



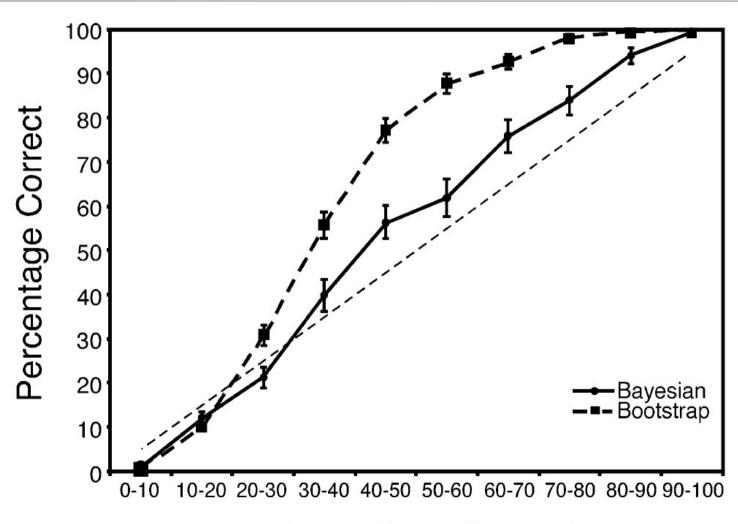
Symphalangus syndactylus
Nomascus concolor
Nomascus gabriellae
Nomascus leucogenys
Hoolock hoolock
Hylobates pileatus
Hylobates agilis
Hylobates moloch
Hylobates muelleri
Hylobates lar



Symphalangus syndactylus
Nomascus concolor
Nomascus gabriellae
Nomascus leucogenys
Hoolock hoolock
Hylobates pileatus
Hylobates agilis
Hylobates moloch
Hylobates muelleri
Hylobates lar



Node support



Percentage Support

 Bayesian model selection is usually based on the marginal probability of the data, conditioned on the model:

Pr(D|M)

- This is a weighted average of the likelihood
- Weights are given by the prior distribution

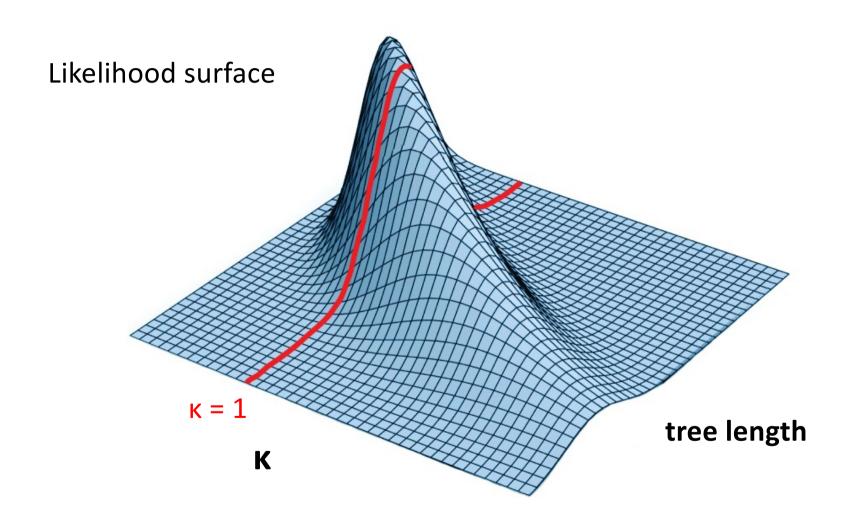
Marginal likelihood of the model

- Compare marginal likelihoods of competing models
- Ratio of marginal likelihoods is the Bayes factor

$$BF = \frac{Pr(D|M_1)}{Pr(D|M_2)}$$

$$logBF = logPr(D|M_1) - logPr(D|M_2)$$

- Models do not need to be nested
- Do not need to correct for number of parameters



Interpreting the Bayes factor

BF	log <i>BF</i>	Evidence against M ₂
1-3	0-1	Not worth mentioning
3 – 20	1-3	Positive
20 – 150	3-5	Strong
> 150	> 5	Very strong

Software for Bayesian Phylogenetics









BEAST 1

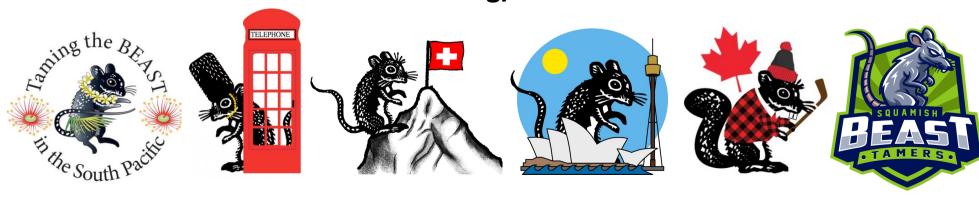
- Bayesian Evolutionary Analysis by Sampling Trees
- Analyse population- or species-level data
- Simultaneous estimation of tree and node times
- Range of clock models
- Range of tree priors and demographic models





- Re-write of BEAST to increase modularity
- Users can extend BEAST by adding packages
- Additional tree priors not available in BEAST 1
- Capacity to perform simulations

For a comparison of *BEAST* 1 and 2: www.beast2.org/beast-features



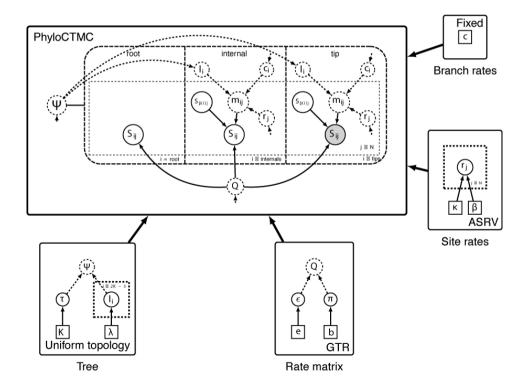
MrBayes

- Primarily designed for species-level data
- Simultaneous estimation of tree and node times
- Range of clock models
- Range of tree priors
- Multiple chains and MCMC diagnostics



RevBayes

- Uses its own R-like language, Rev
- Interactive construction of graphical model
- Flexible and can be used for simulation and inference





Useful references

