



Taming the BEAST

Bayesian Evolutionary Analyses by Sampling Trees

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Setting priors

Some slides inspired by slides from
Veronika Bošková, Jana
Huisman, Vladimir Minin and
Chi Zhang

A diagram illustrating Bayes' theorem. The equation is $P(\text{model} \mid \text{data}) = \frac{P(\text{data} \mid \text{model})P(\text{model})}{P(\text{data})}$. Handwritten red arrows and labels identify the components: 'Likelihood' points to $P(\text{data} \mid \text{model})$, 'Prior' points to $P(\text{model})$, 'Posterior' points to $P(\text{model} \mid \text{data})$, and 'Model evidence' points to $P(\text{data})$.

$$P(\text{model} \mid \text{data}) = \frac{P(\text{data} \mid \text{model})P(\text{model})}{P(\text{data})}$$

Likelihood

Prior

Posterior

Model evidence

Prior → P(model)

- Original probability for the model parameters/components (before data collected or analysed)
- Belief in our hypothesis
- All parameters have priors, whether you specify them or not!

What is a **prior**?

- Usually a **distribution** for some model **parameter**
e.g. clock rate, tree, ...
- Often a **parametric distribution**
e.g. uniform, normal, gamma, beta, lognormal, Laplace, ...
- Sometimes a **prior** on a model component
e.g. substitution model (HKY, GTR, JC, ...)
- Priors can have priors which can have priors *ad infinitum*
(hyperpriors)
- Parameter bounds are part of the prior
e.g. normal distribution with lower bound 0.

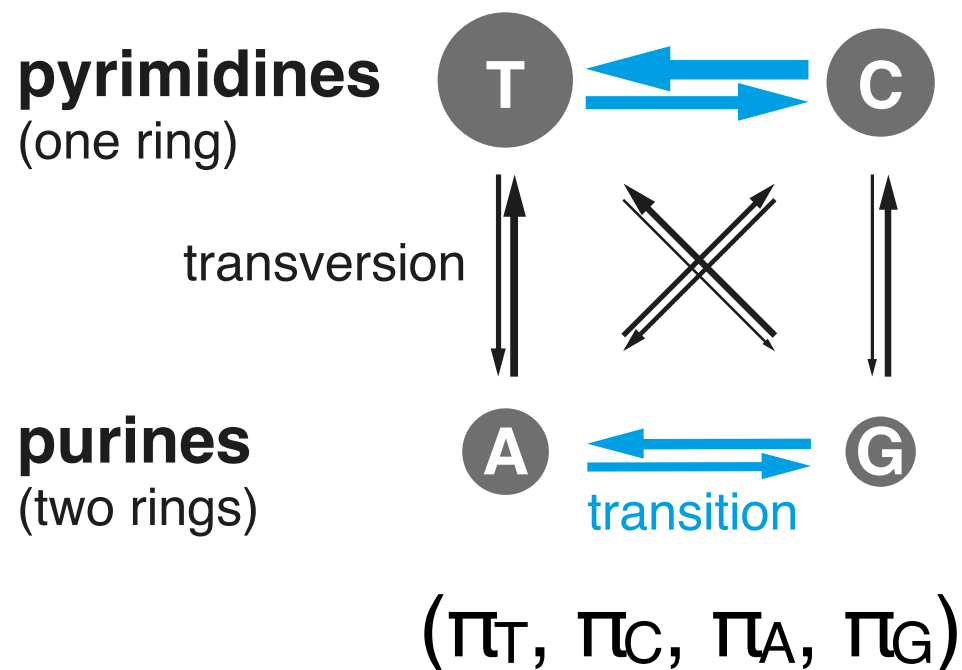
How to pick a **prior**?

- Should be chosen based on your beliefs about model parameters (from independent evidence/experiments)
- Should be chosen with a particular analysis in mind (no priors are universal!)
- Think carefully about plausible ranges of parameters
- Be conservative if you are unsure about the parameter (use diffuse priors)
- Conjugate priors?
(no — those are just for mathematical convenience)
- Reference priors?
(yes — if you are an objective Bayesian)
- Do not use improper priors if you can help it
(priors that integrate to infinity, e.g. $1/x$)





Example: HKY-model (HKY85)



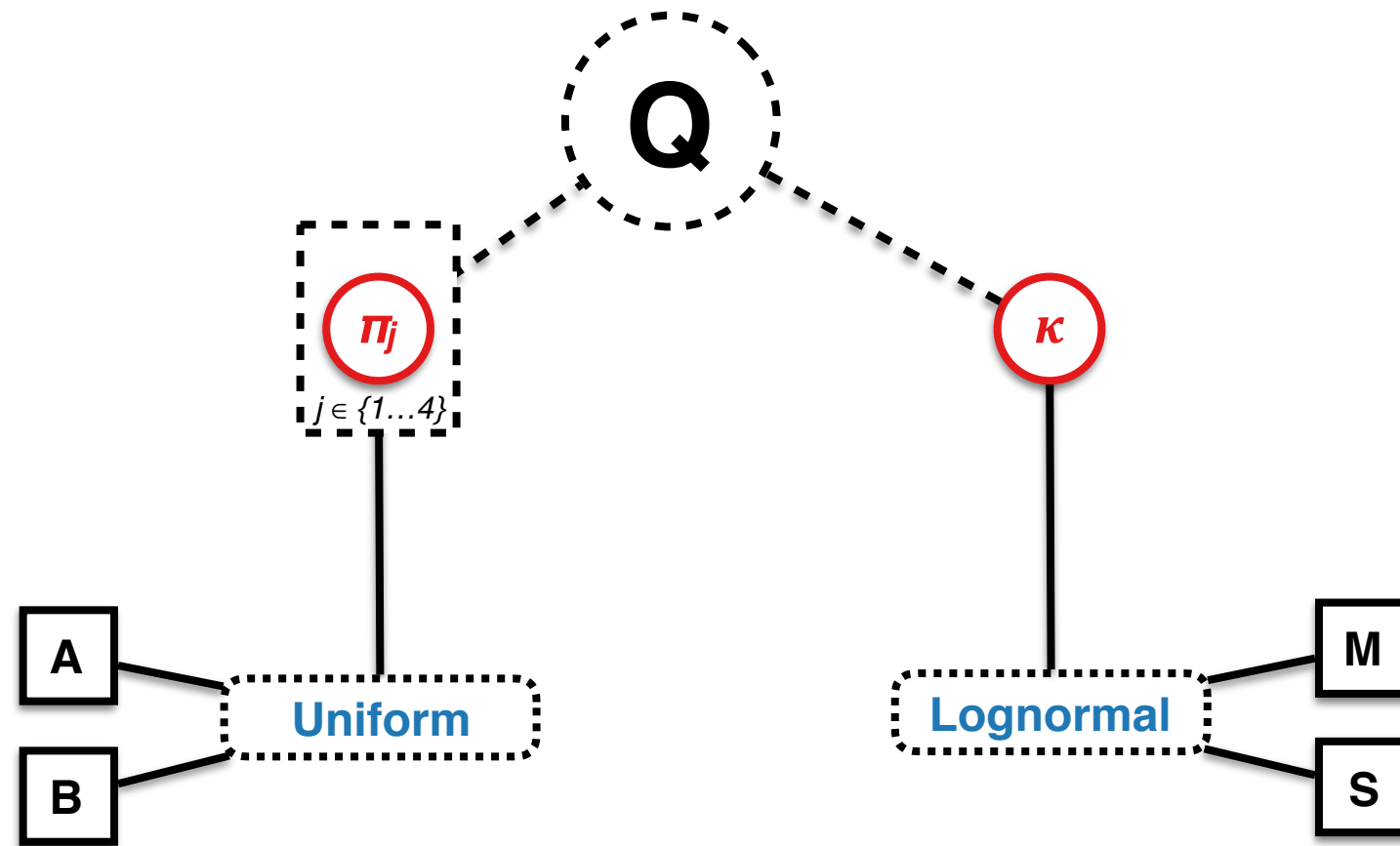
$$\begin{matrix} & T & C & A & G \\ \begin{matrix} T \\ C \\ A \\ G \end{matrix} & \begin{pmatrix} \cdot & \alpha\pi_C & \beta\pi_A & \beta\pi_G \\ \alpha\pi_T & \cdot & \beta\pi_A & \beta\pi_G \\ \beta\pi_T & \beta\pi_C & \cdot & \alpha\pi_G \\ \beta\pi_T & \beta\pi_C & \alpha\pi_A & \cdot \end{pmatrix} \end{matrix}$$

$$= \begin{pmatrix} \cdot & \alpha & \beta & \beta \\ \alpha & \cdot & \beta & \beta \\ \beta & \beta & \cdot & \alpha \\ \beta & \beta & \alpha & \cdot \end{pmatrix} \cdot \begin{pmatrix} \pi_T & 0 & 0 & 0 \\ 0 & \pi_C & 0 & 0 \\ 0 & 0 & \pi_A & 0 \\ 0 & 0 & 0 & \pi_G \end{pmatrix}$$

- **Q** matrix gives relative rates of substitution between nucleotides
- 5 parameters:
 - $\kappa = \alpha/\beta$
 - $\pi_T, \pi_C, \pi_A, \pi_G$ — assume these all have the same prior probability

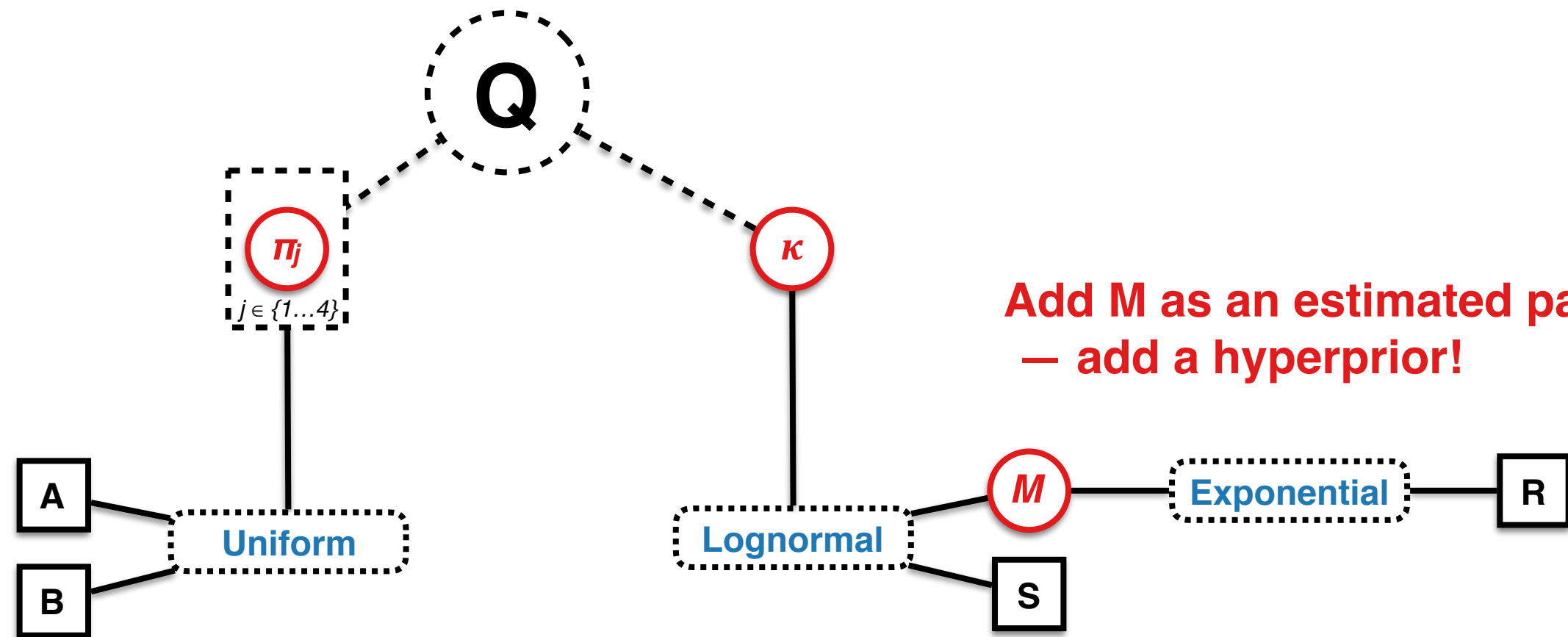


Example: HKY-model (HKY85)





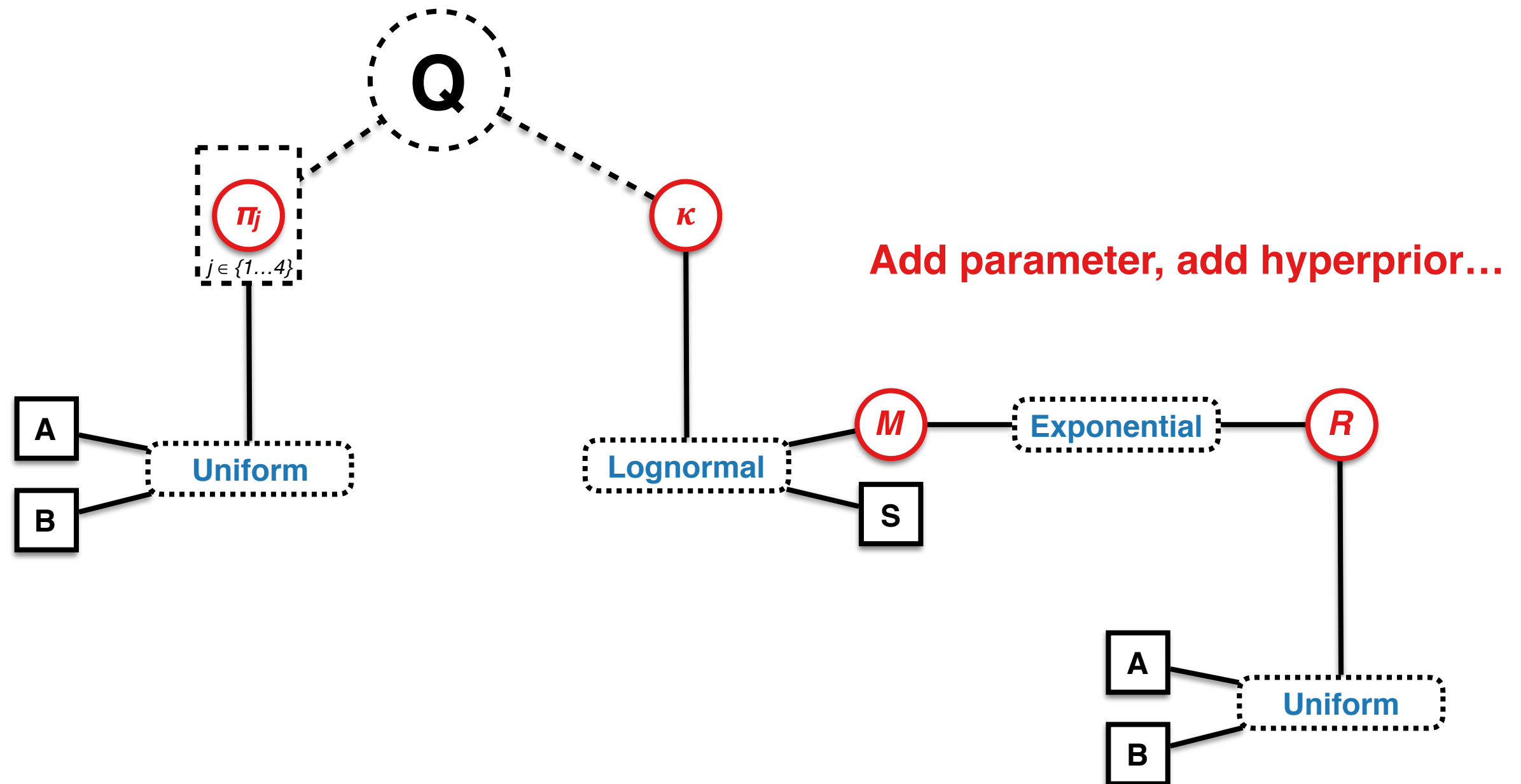
Example: HKY-model (HKY85)



Add M as an estimated parameter
— add a hyperprior!

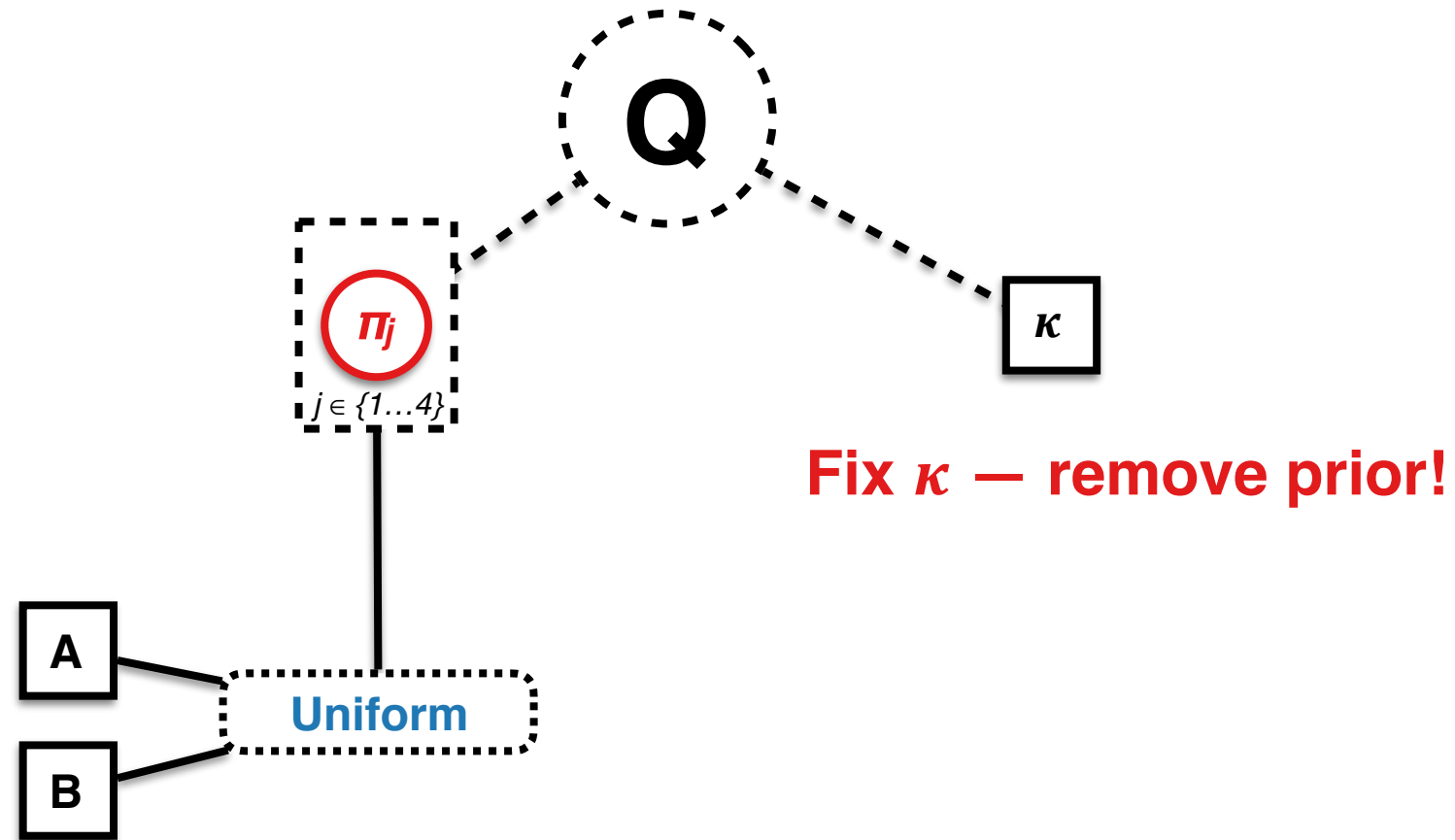


Example: HKY-model (HKY85)

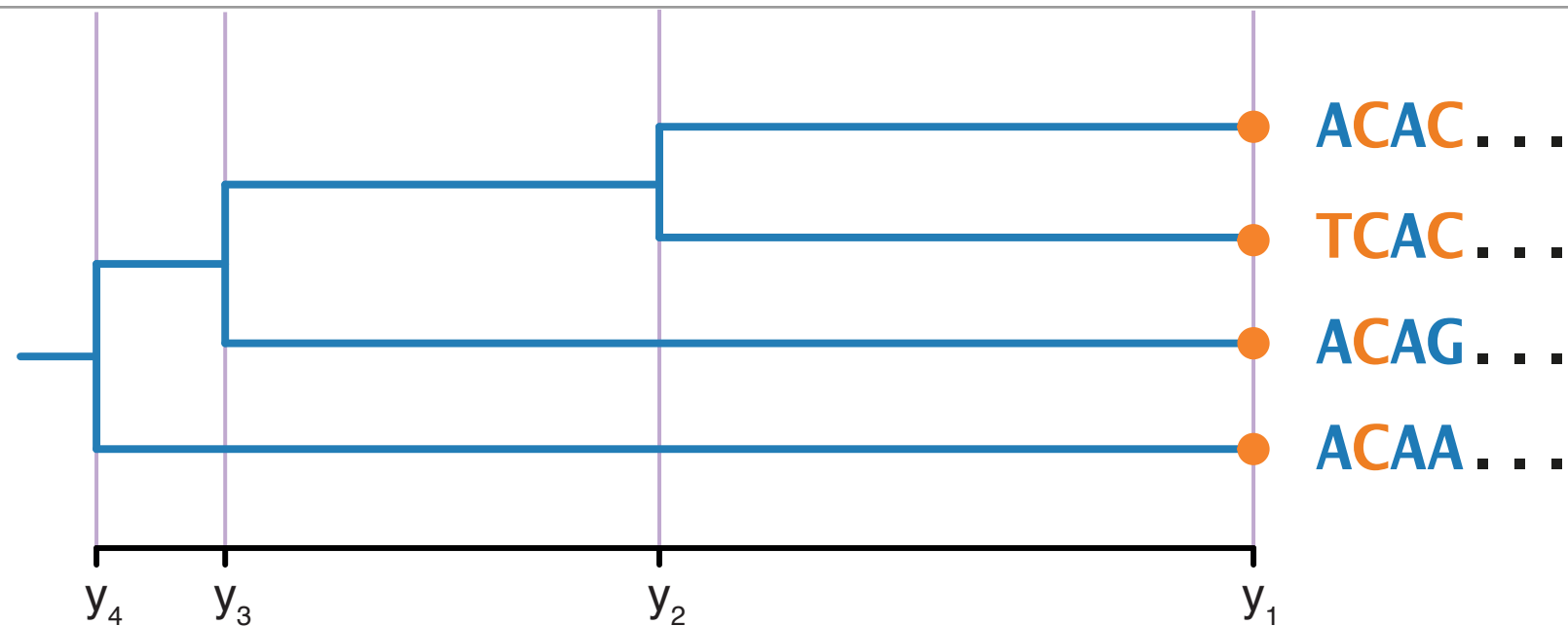




Example: HKY-model (HKY85)



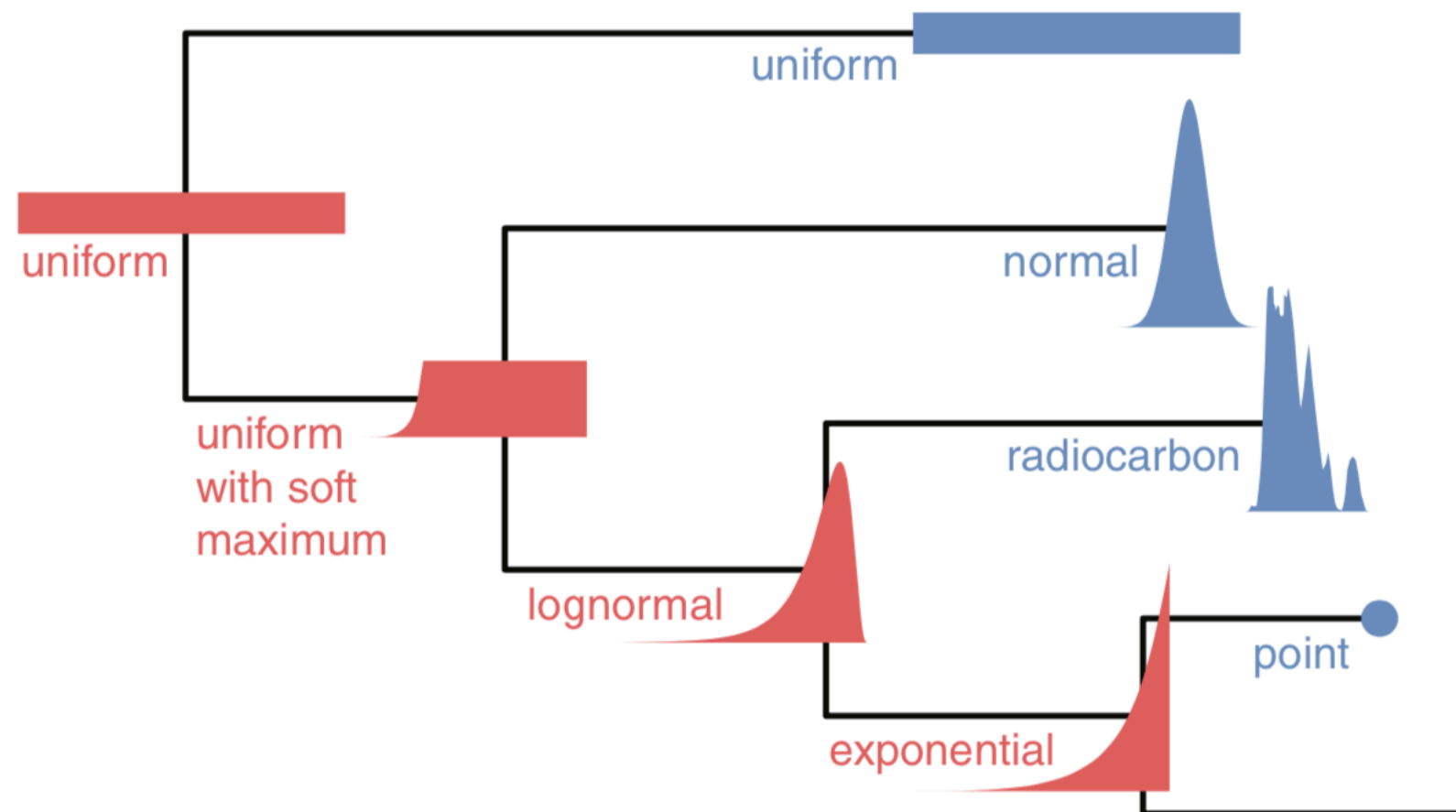
Calibration nodes



Homochronous trees need external **(prior)** information to date:

- Fix the clock rate
- Use a **calibration node** (or nodes)

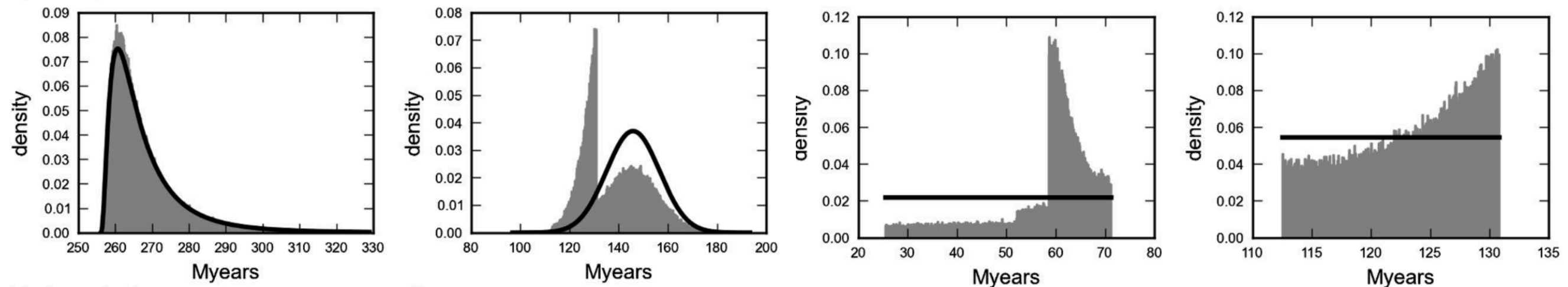
Calibration nodes



Calibration nodes

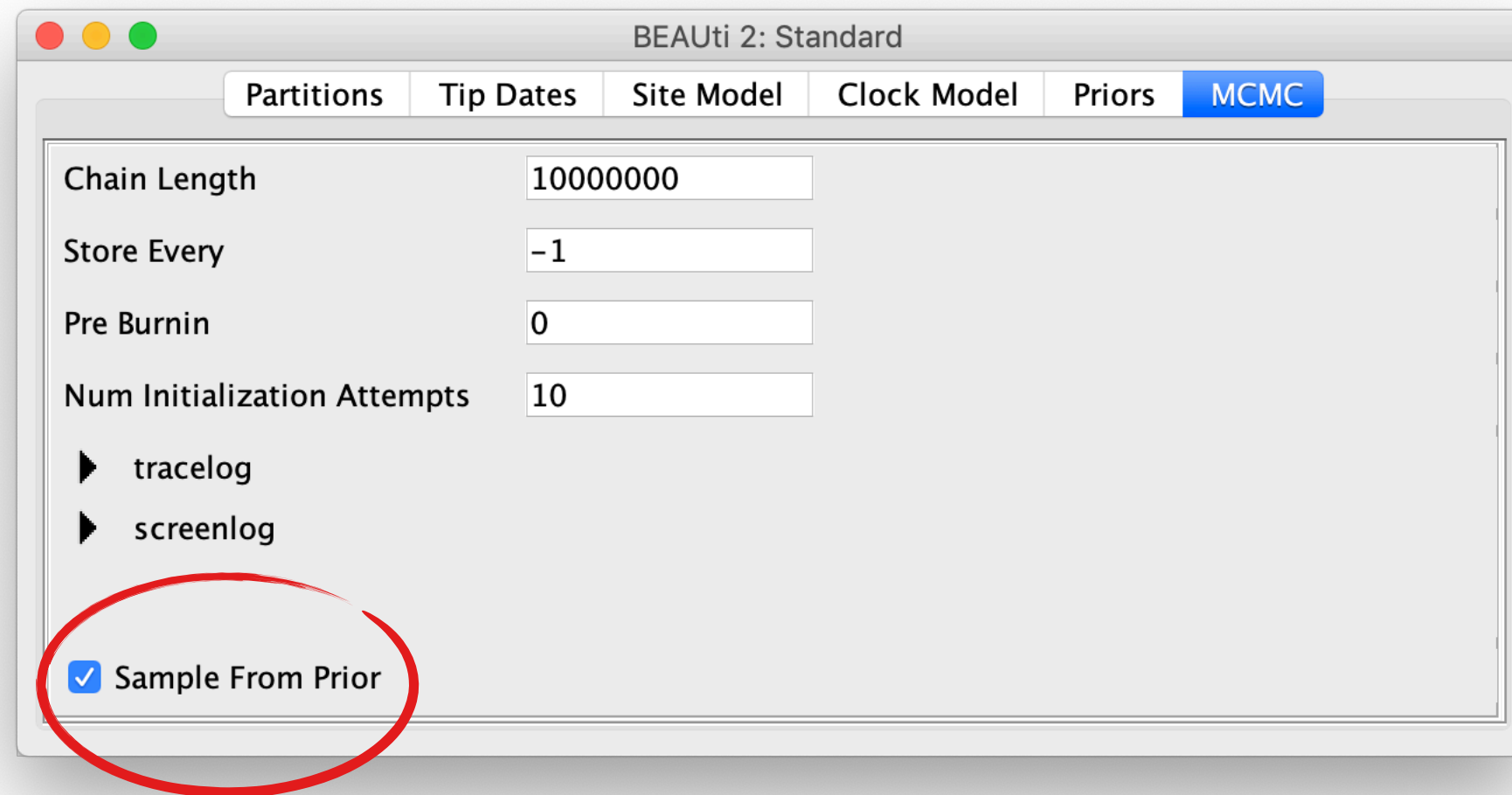
More calibration nodes mean more information to calibrate the clock

but is more better?



Induced prior may be different to the prior you set!

Setting priors best practice



- Sample or simulate from prior
- For key distributions plot prior and posterior together
- Mess with priors and see how sensitive/robust posterior is to prior perturbations