

RevBayes: Bayesian phylogenetics using probabilistic graphical models and an interpreted model specification language

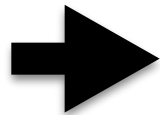
SEBASTIAN HÖHNA

**DEPARTMENT OF EVOLUTION & ECOLOGY, UC DAVIS
DEPARTMENT OF STATISTICS, UC BERKELEY**

RevBayes

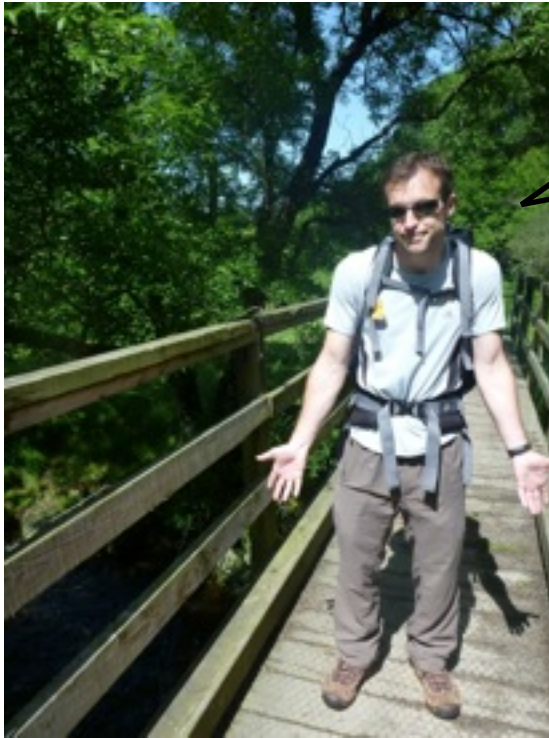
- Why a new software???

- No inherited problems
- Extendable software
- We need a general a flexible modeling framework
- It needs to be fast
- We need to be comfortable to develop in it



**A new software provides more opportunities
but is also more work.**

Ambitions



The limit is the sky!

Bastien Boussau

Ambitions

Your wildest dreams
will come true!



Fredrik Ronquist

History

2004, Work starts with NIH grant to J.P. Huelsenbeck, F. Ronquist and B. Larget. P. van der Mark starts PostDoc.

Spring 2005, Meeting in San Diego about goals and key concepts (J.P. Huelsenbeck, F. Ronquist, B. Larget. P. van der Mark and Donald Simon).
Codename: MrBayes 4

August 2009, Sourceforge repository created and first lines written.

August 2011, Developers workshop in Berkeley.

March 2013, Alpha testing during a workshop in Groningen.

2004

May 2005, Release of MrBayes 3.1

2006

Summer 2005 and early 2006, next meetings (now including M. Suchard). Focus shifts to R-like model specification

2008

April 2009, S. Höhna starts PhD and joint the project

2010

June 2011, Rewrite of RevBayes

2012

June 2012, Rewrite of RevBayes

2014

August 2014, First official user workshop at NESCent. RevBayes is ready at a beta stage.

Aims for RevBayes



- 1) General and flexible model specification
 - a) Availability of (common) models
 - b) Extendability

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2) Easy to learn

- a) Well structured model specification
- b) Explicit models
- c) Documentation, examples and tutorials

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3) Fast & Efficient

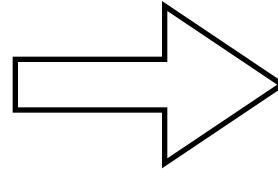
- a) Fast likelihood calculators
- b) Efficient (MCMC) algorithms, e.g., tree proposals

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**Graphical
Models
&
interpreted
interface
&
C++**

Specifying a model in MrBayes

Model settings for partition 1:

Parameter	Options	Current Setting
Nucmodel	4by4/Doublet/Codon	4by4
Nst	1/2/6	6
Code	Universal/Vertmt/Mycoplasma/ Yeast/Ciliates/Metmt	Universal
Ploidy	Haploid/Diploid	Diploid
Rates	Equal/Gamma/Propinv/Invgamma/Adgamma	Invgamma
Ngammacat	<number>	4
Nbetacat	<number>	5
Omegavar	Equal/Ny98/M3	Equal
Covarion	No/Yes	No
Coding	All/Variable/Noabsencesites/ Nopresencesites	All
Parsmodel	No/Yes	No



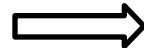
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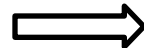
...



HKY+G - Substitution Model



...



4-Gamma Categories



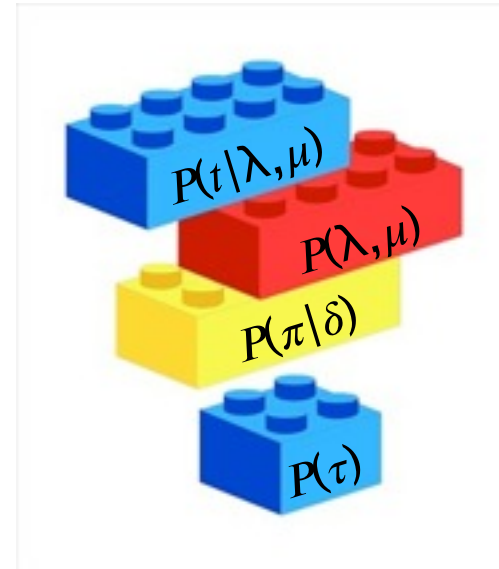
...



Constant-Clock Model

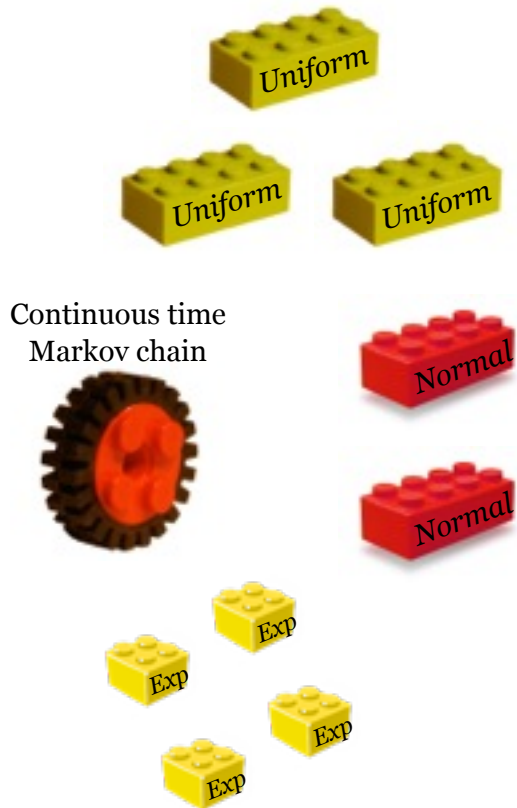
Bottom Up Design

- Standard building blocks:
 - Distributions:
 - Uniform Distribution
 - Normal Distribution
 - Exponential Distribution
 - Gamma Distribution
 - ...
 - Functions:
 - Addition
 - Multiplication
 - Exponentiation
 - ...
- Phylogenetically inspired
 - Distributions
 - Tree priors
 - Substitution processes
 - ...
 - Functions
 - Rate matrix
 - ...
- Assemble model from probability distributions and functions



Bottom Up Design

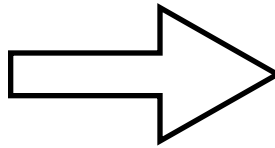
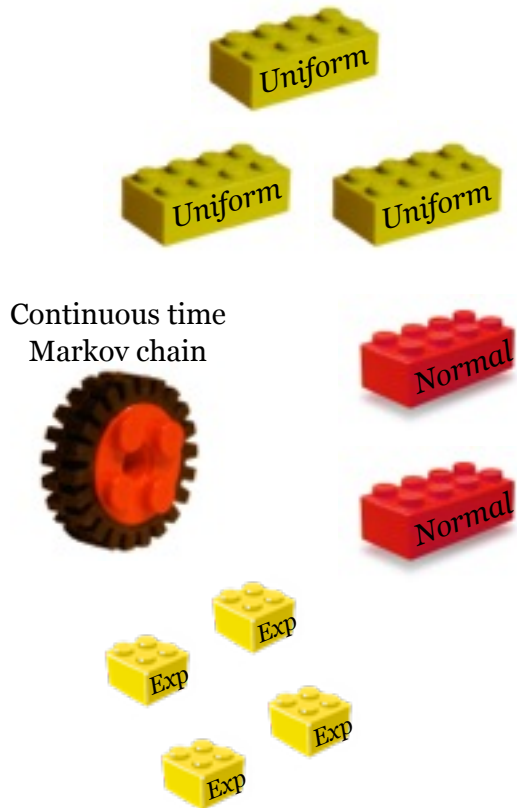
Building blocks:



Assembled model:

Bottom Up Design

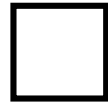
Building blocks:



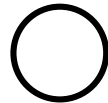
Assembled model:



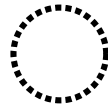
Graphical Models Notation



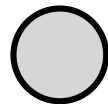
a) Constant node



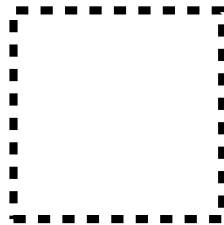
b) Stochastic node



c) Deterministic node

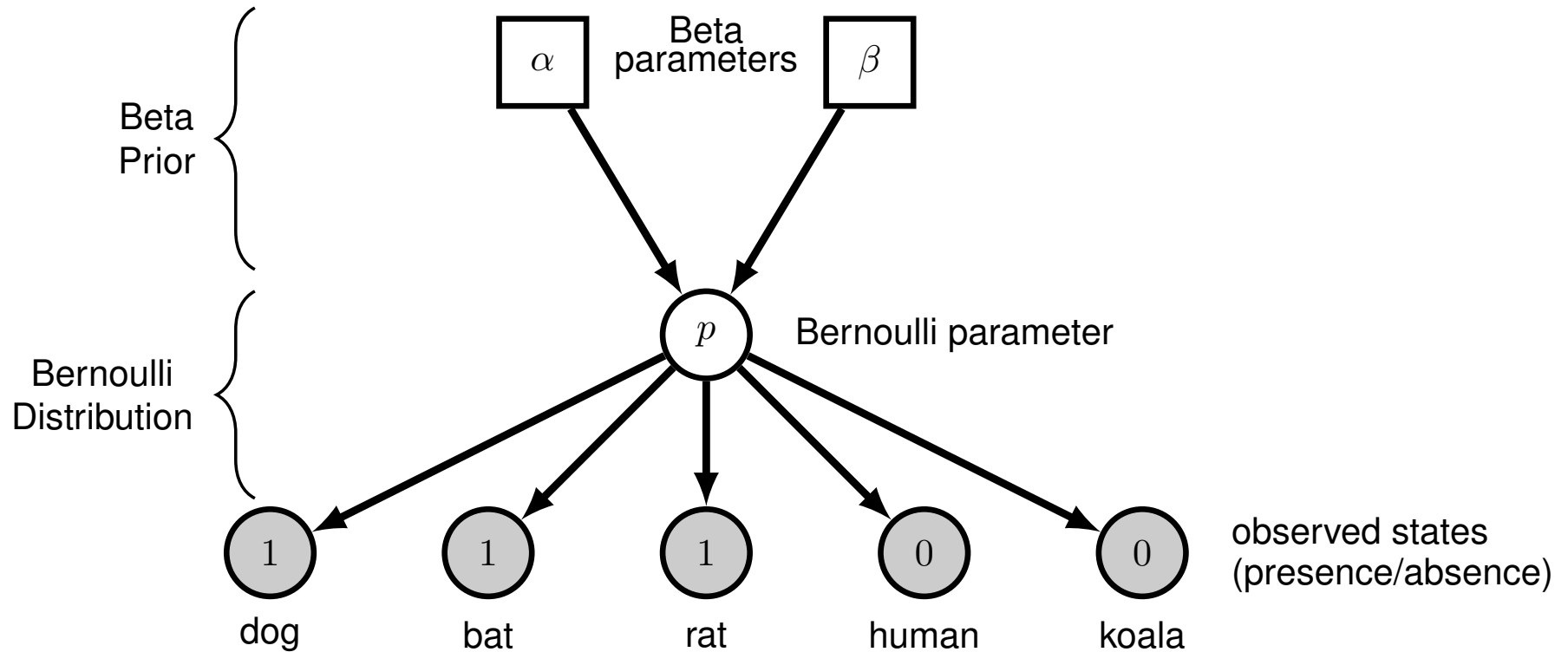


d) Clamped node
(observed)

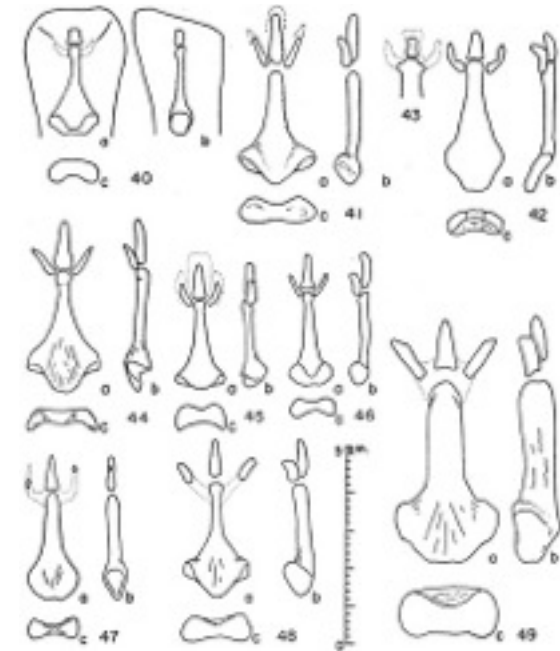
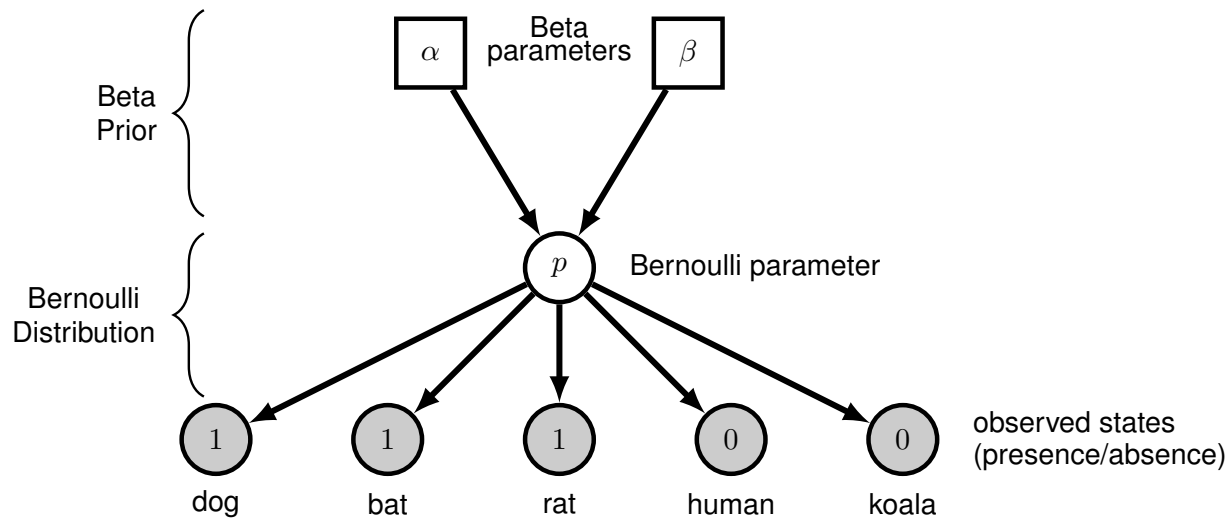


e) Plate

A simple graphical model

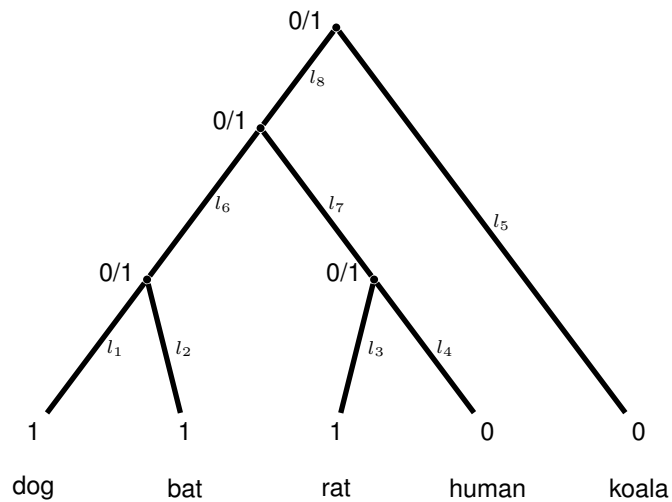


A simple graphical model



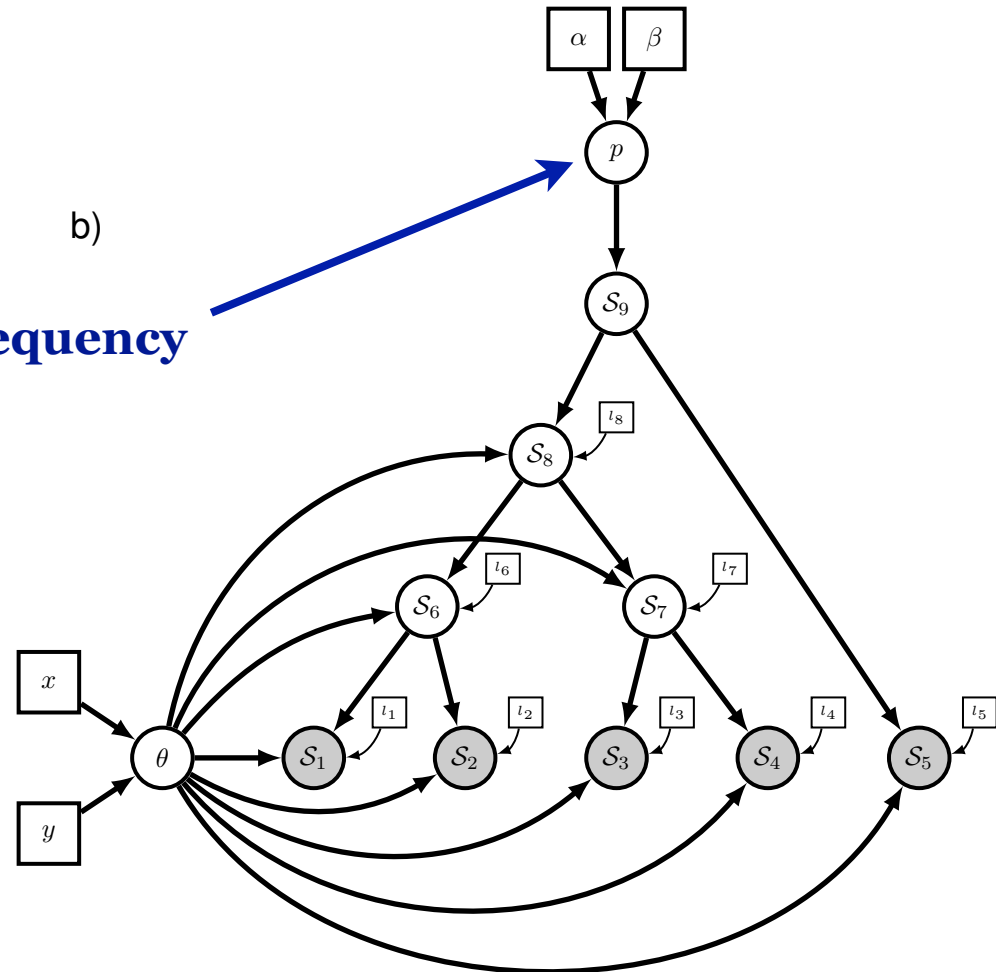
A graphical tree model

a)

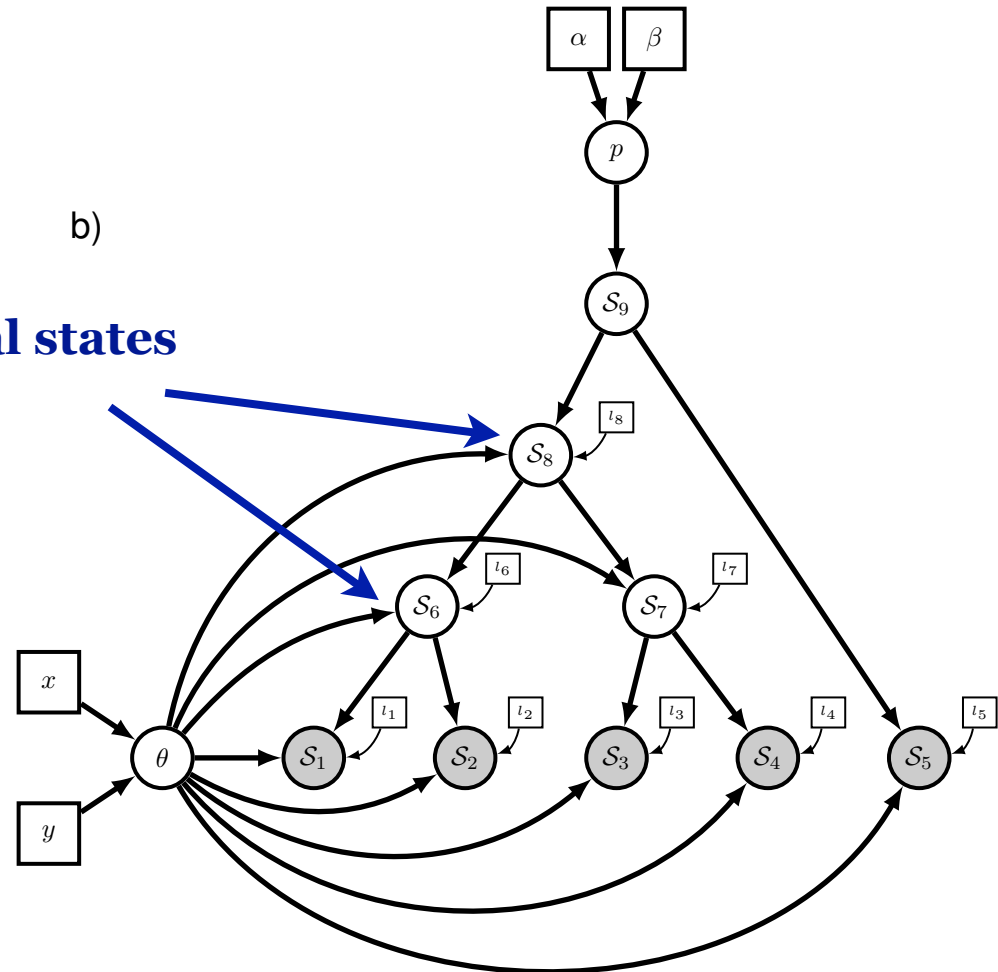


b)

root frequency

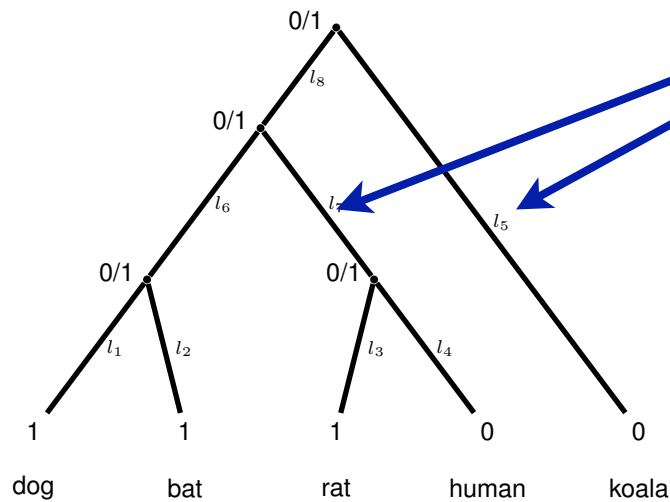


internal states

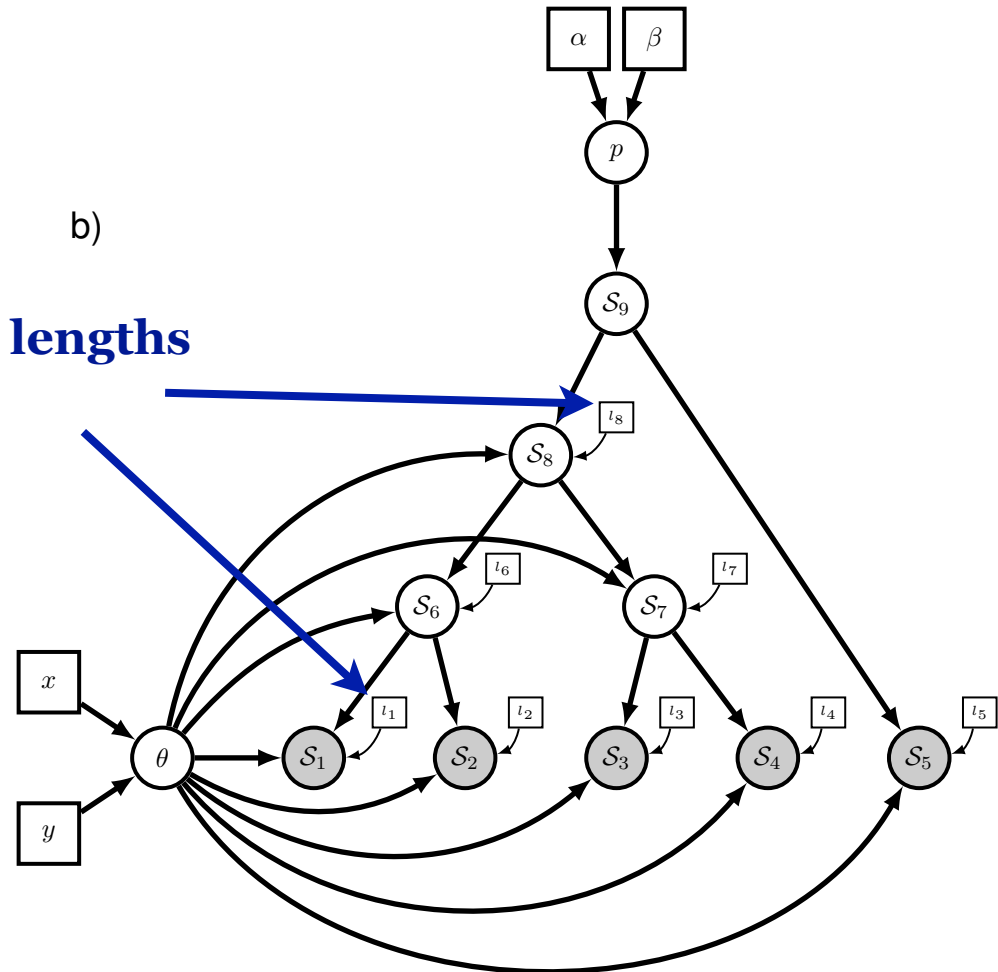


A graphical tree model

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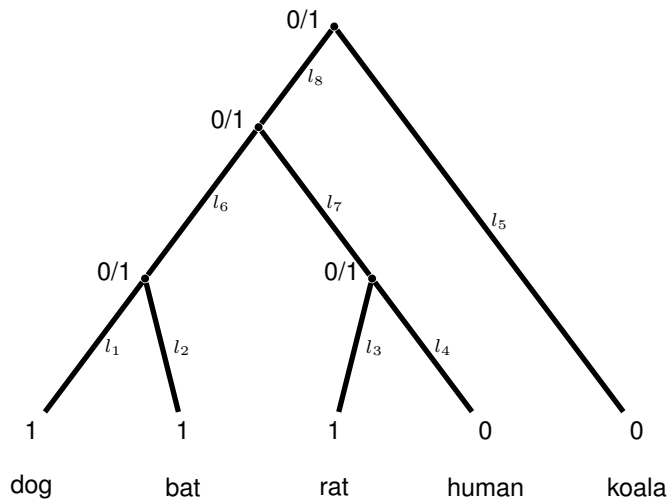


b)



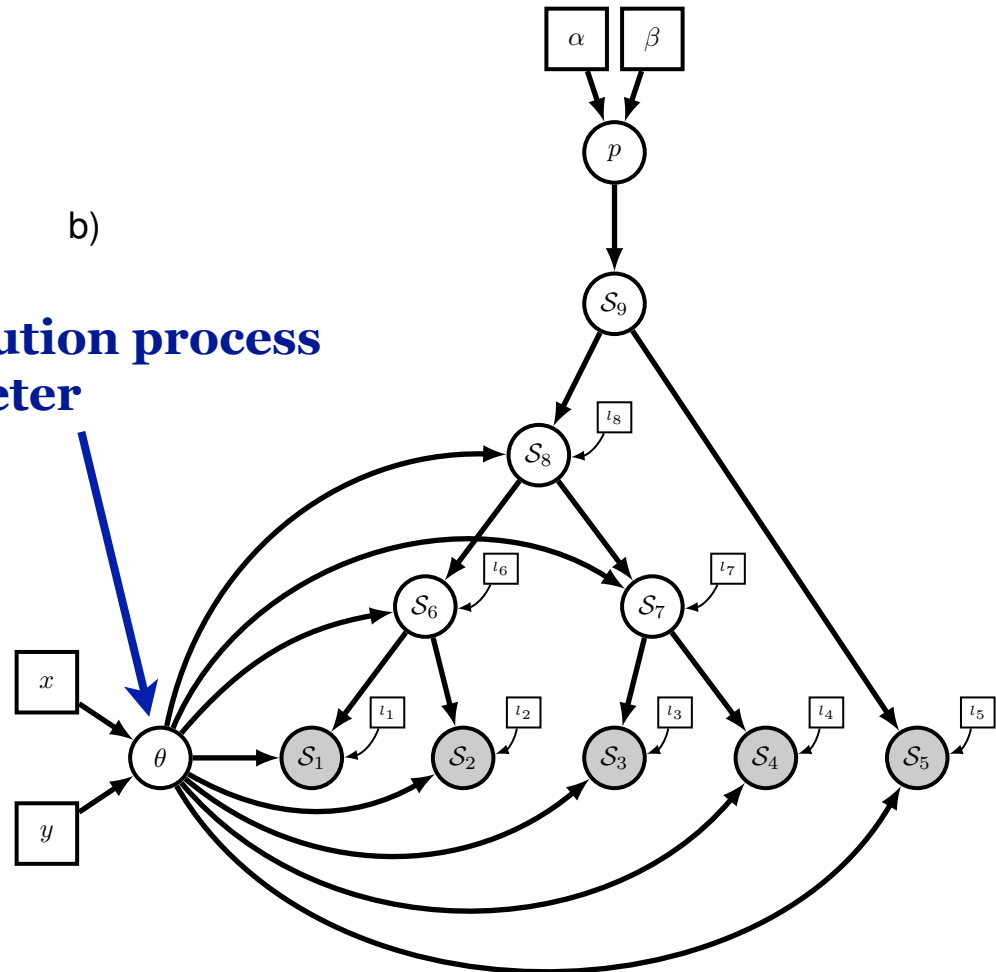
A graphical tree model

a)



b)

**substitution process
parameter**

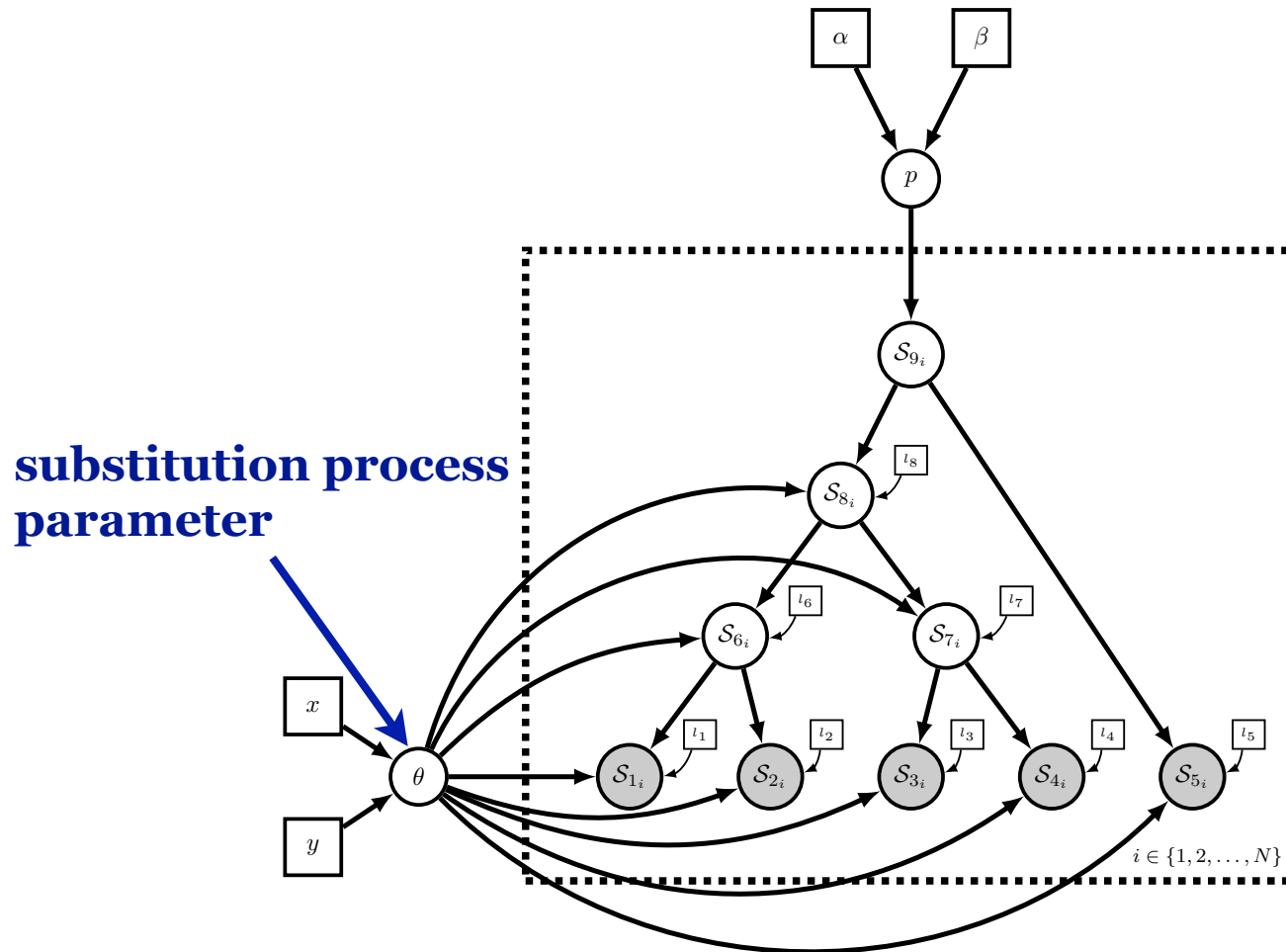


b)

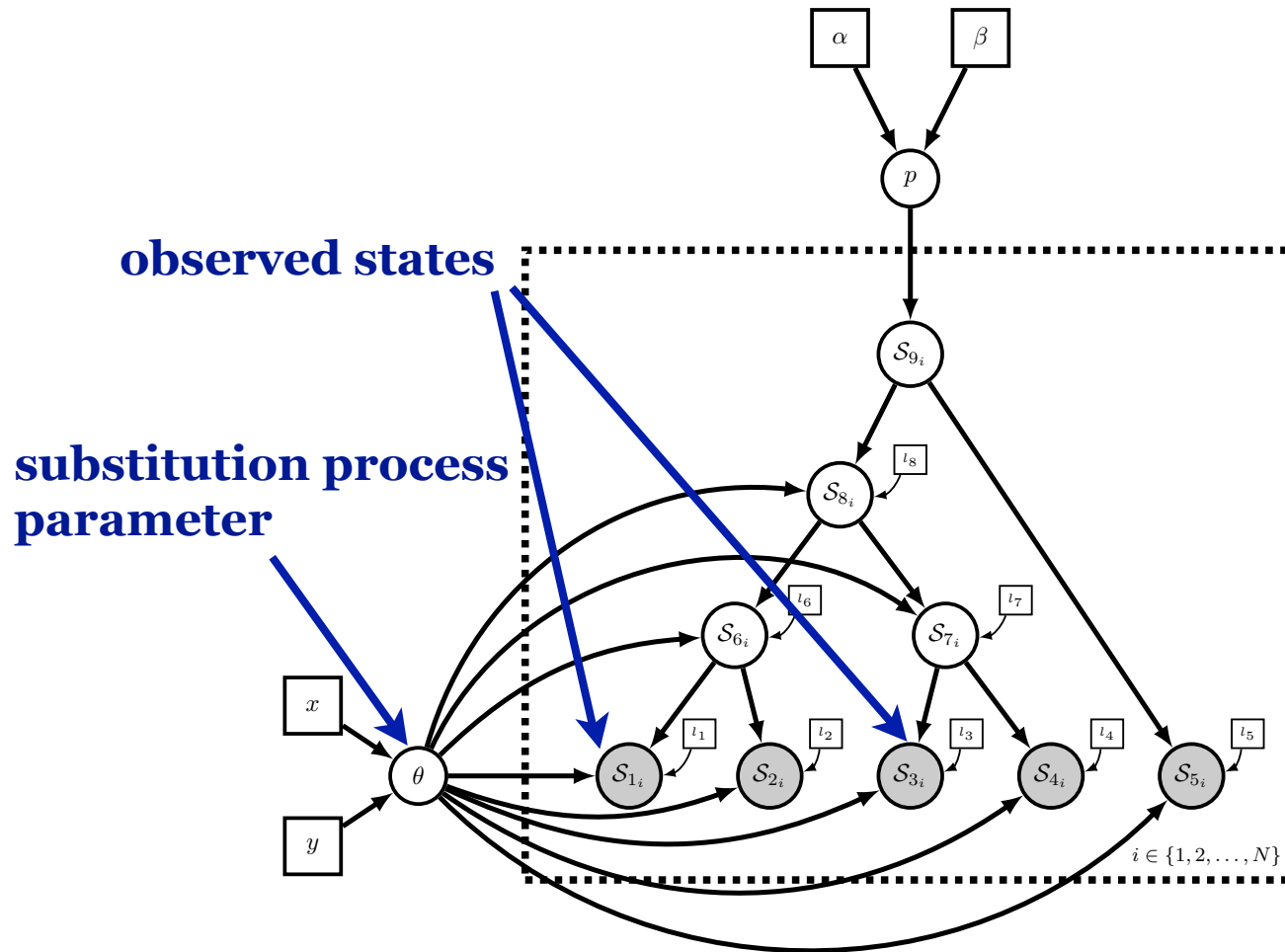
observed states

Diagram illustrating a directed graphical model (Bayesian network) for a sequence of states. The model includes a parameter node θ , a sequence of state nodes s_1, s_2, s_3, s_4, s_5 , and a sequence of latent variable nodes l_1, l_2, l_3, l_4, l_5 . The state nodes s_1 through s_5 are shaded gray, indicating they are observed states. The latent variable nodes l_1 through l_5 are white, indicating they are unobserved. The parameter node θ is white. The model is defined by the following directed edges: $\theta \rightarrow s_1, s_2, s_3, s_4, s_5$; $s_1 \rightarrow s_2, s_3, s_4, s_5$; $s_2 \rightarrow s_3, s_4, s_5$; $s_3 \rightarrow s_4, s_5$; $s_4 \rightarrow s_5$; $s_5 \rightarrow s_8$; $s_8 \rightarrow s_6, s_7$; $s_6 \rightarrow s_1, s_2, s_3, s_4, s_5$; $s_7 \rightarrow s_1, s_2, s_3, s_4, s_5$; $s_1 \rightarrow l_1$; $s_2 \rightarrow l_2$; $s_3 \rightarrow l_3$; $s_4 \rightarrow l_4$; $s_5 \rightarrow l_5$; $s_6 \rightarrow l_6$; $s_7 \rightarrow l_7$; $s_8 \rightarrow l_8$. Two blue arrows point from the text "observed states" to the shaded state nodes s_1 and s_3 .

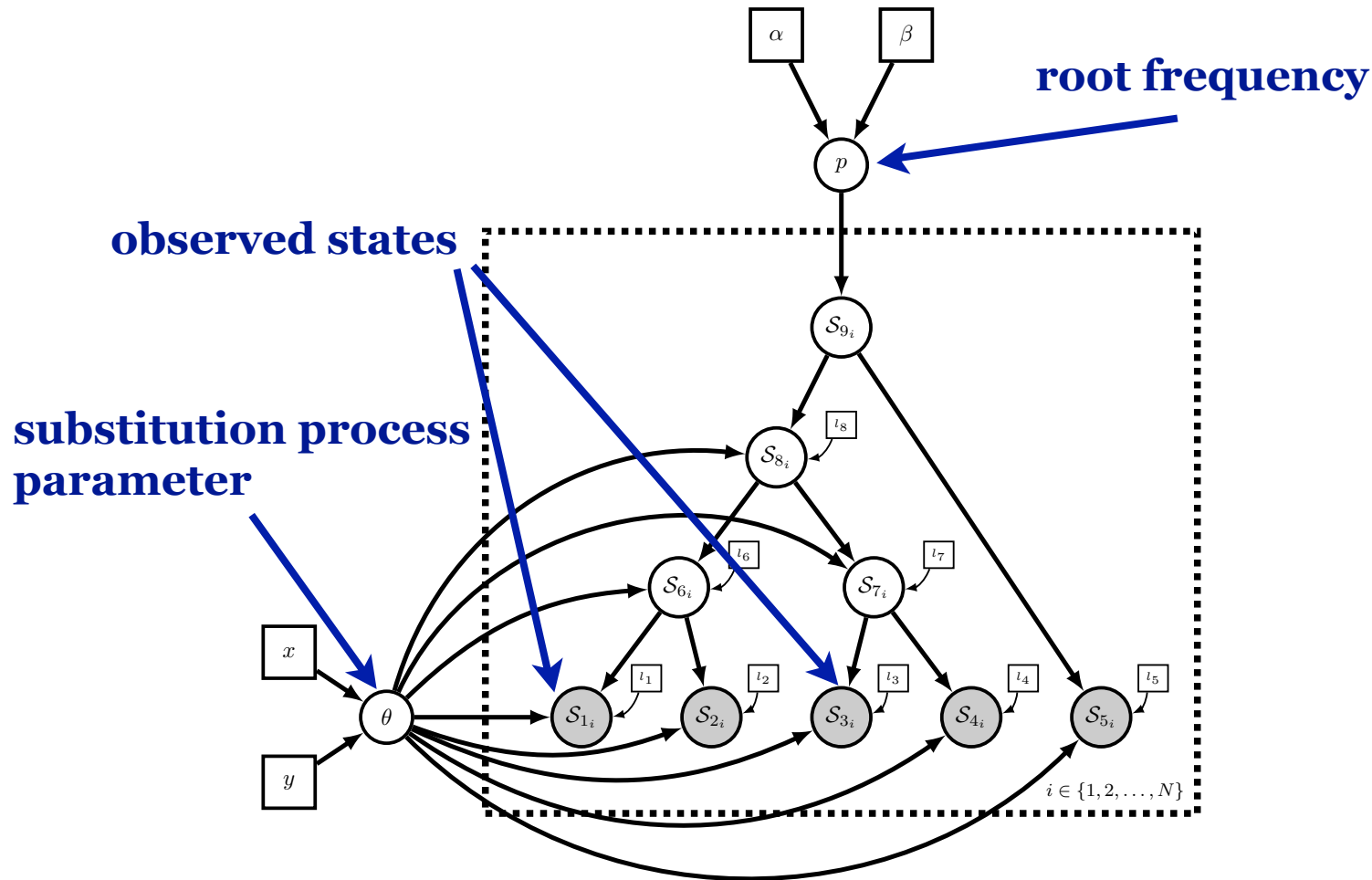
A graphical model for discrete characters



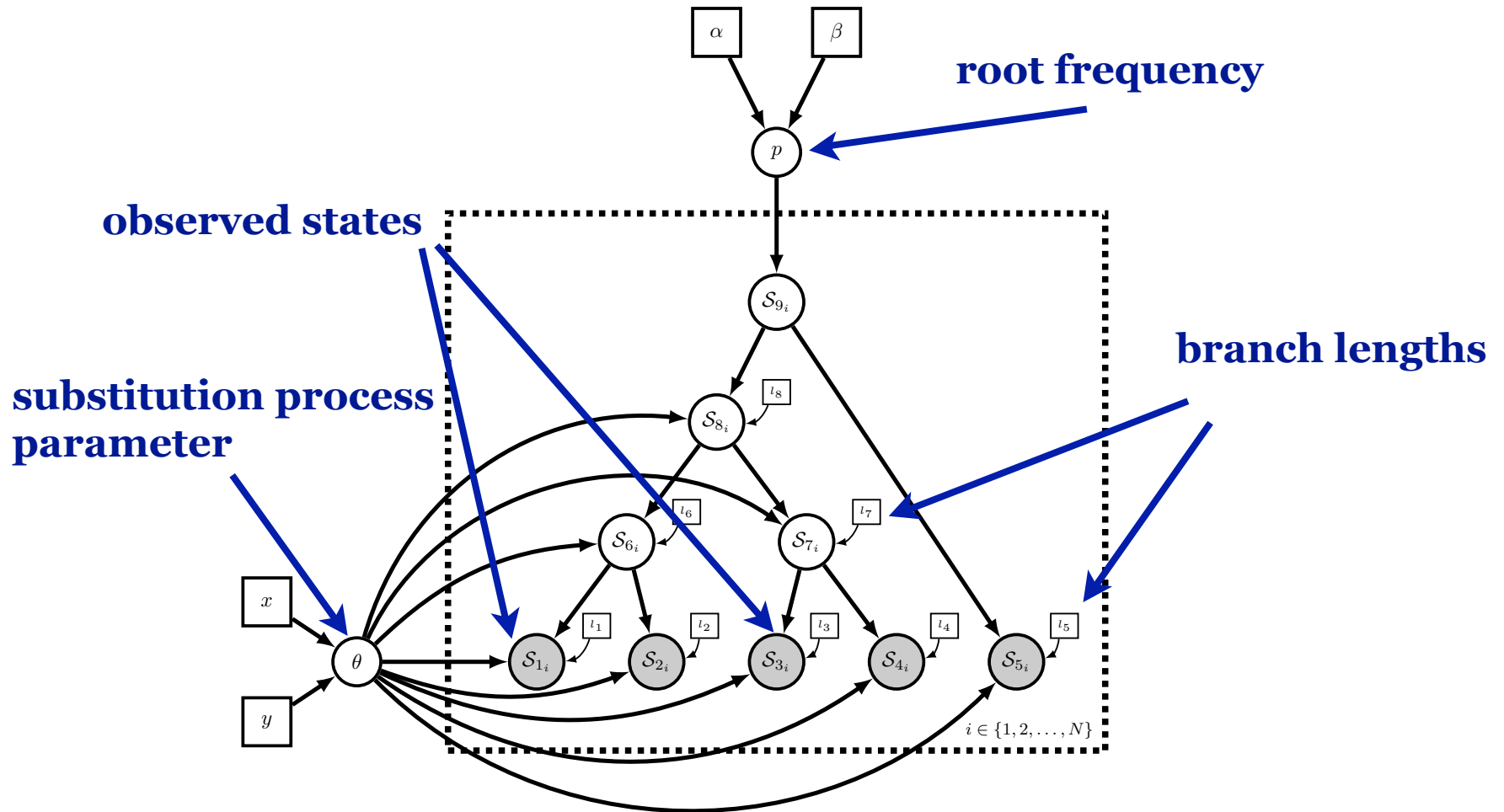
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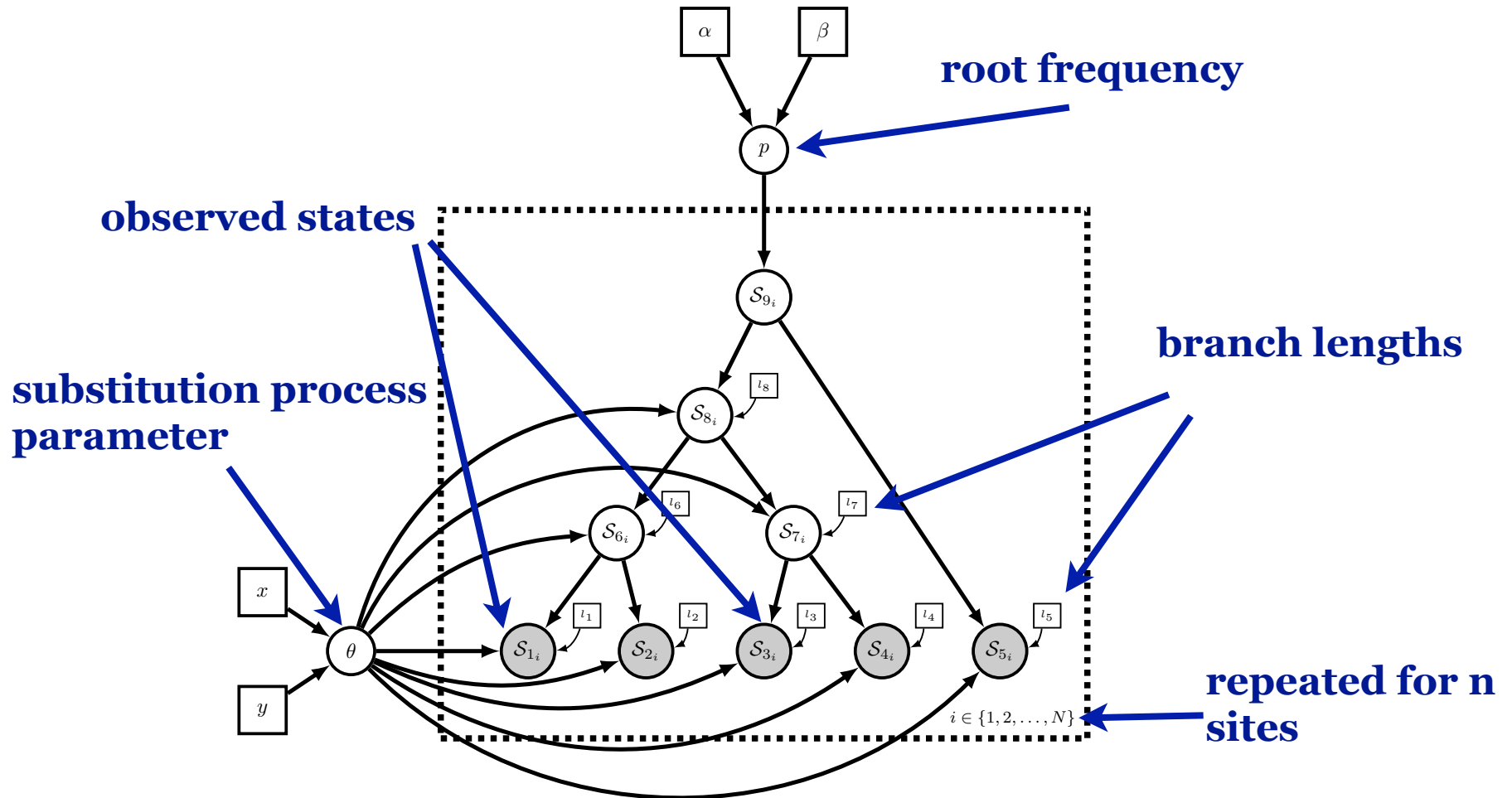
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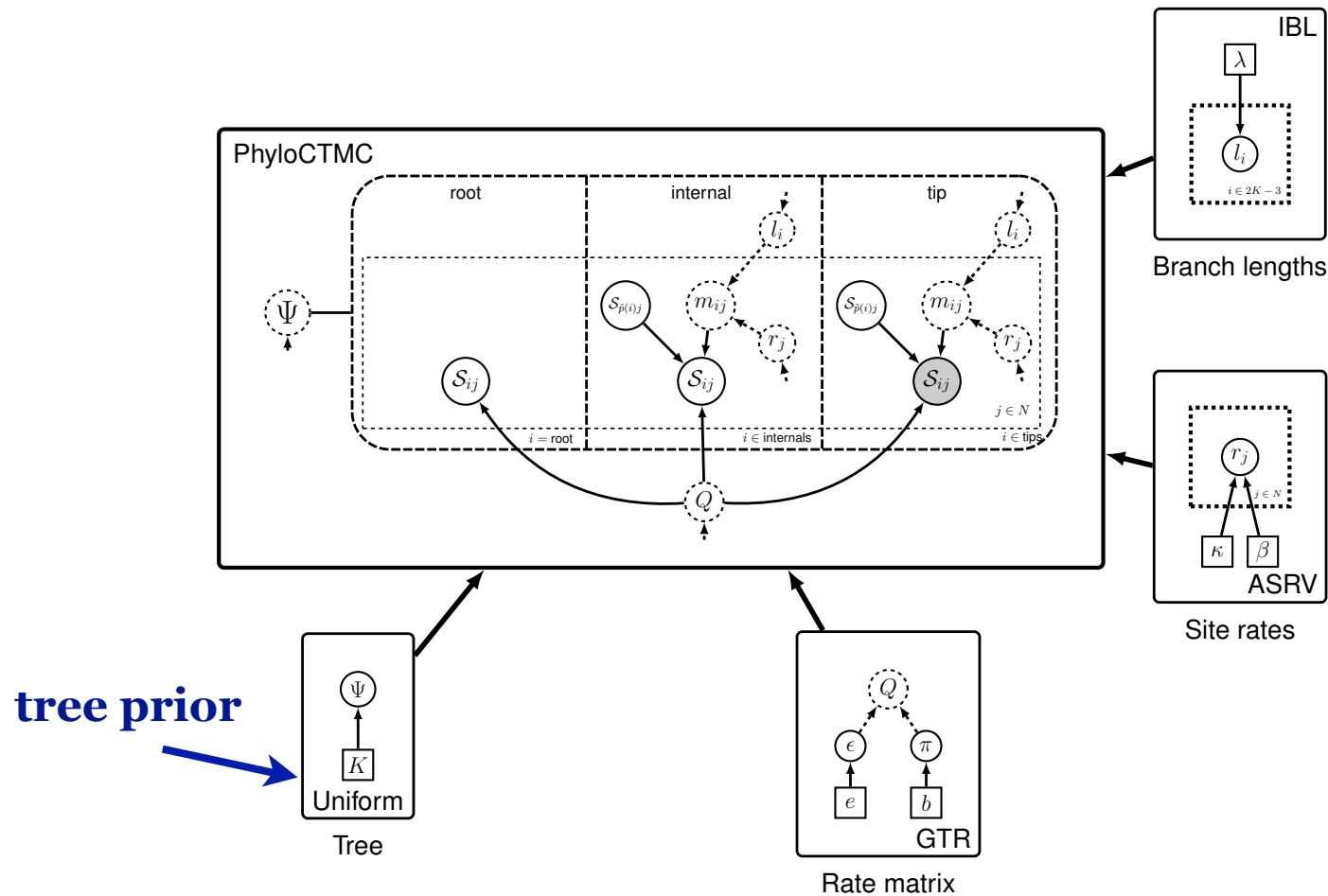
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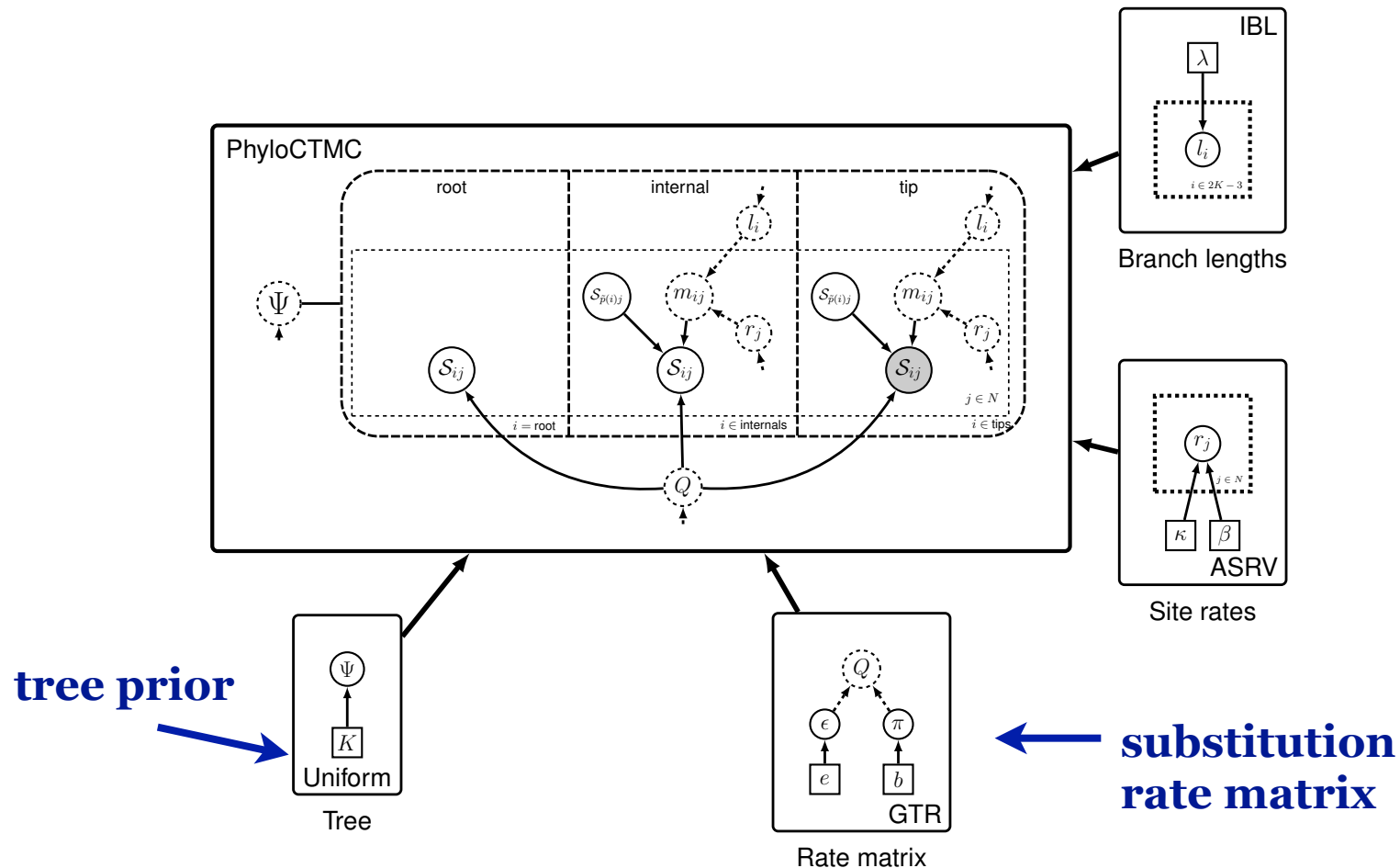
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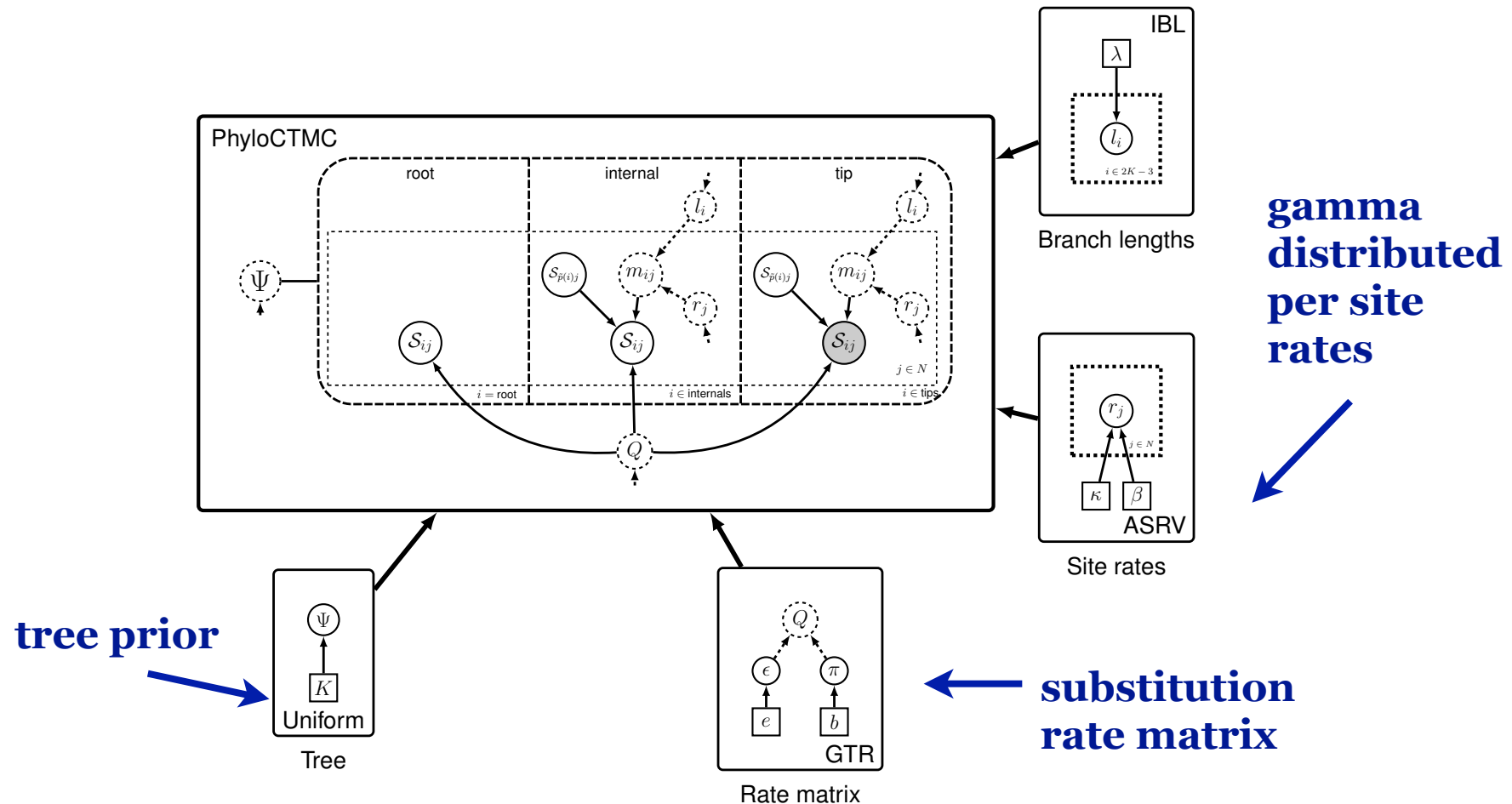
A compact/modular representation



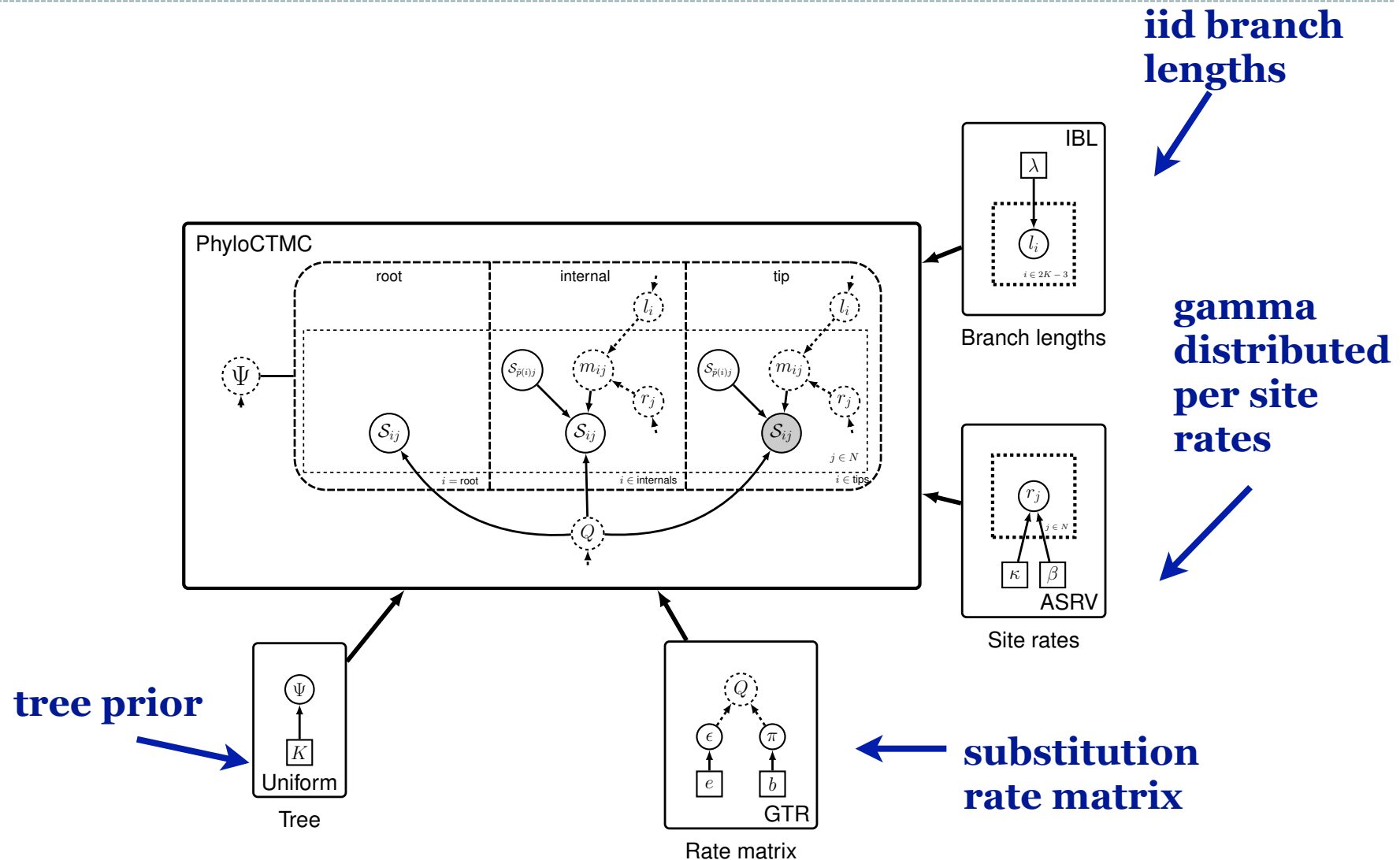
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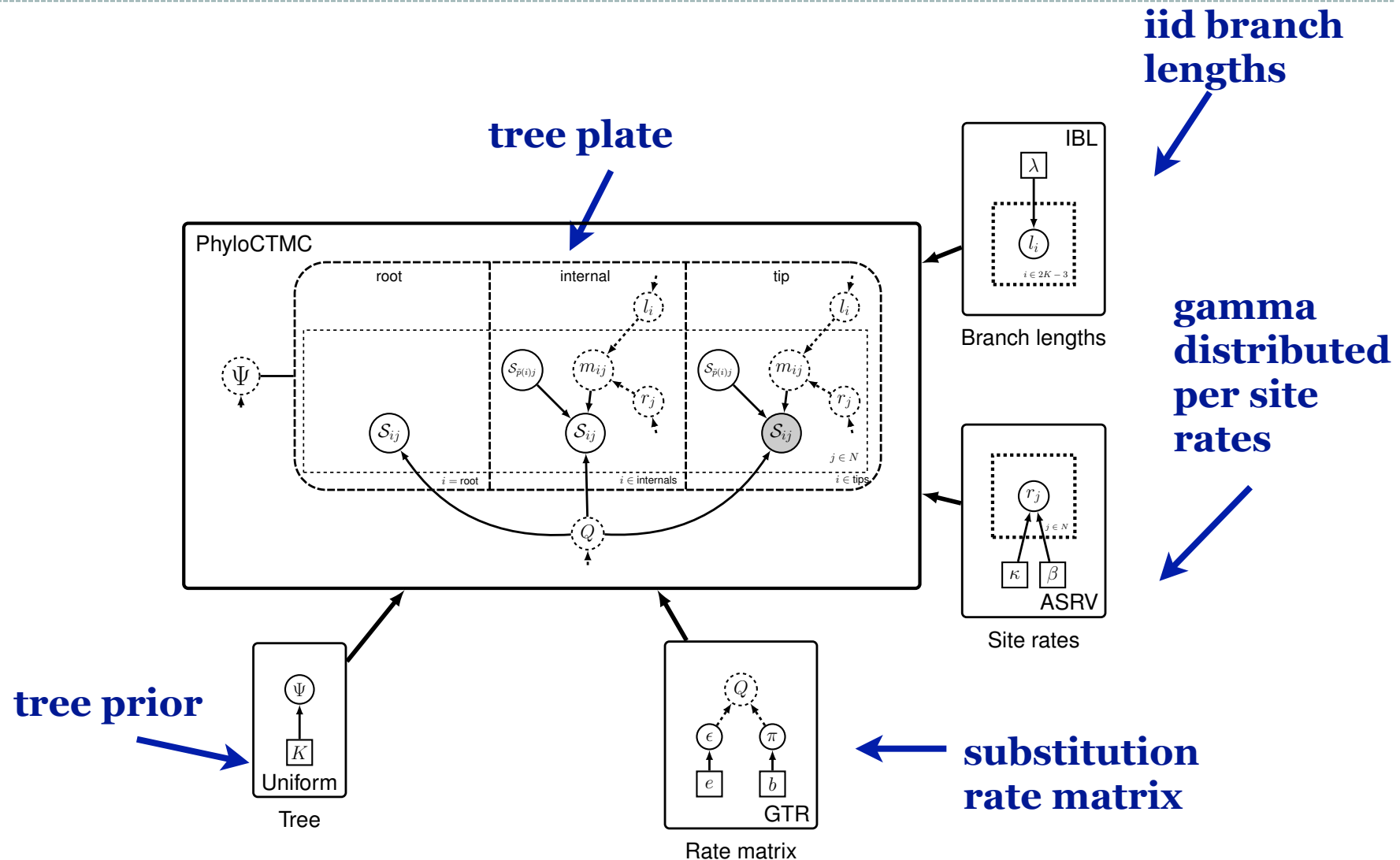
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A compact/modular representation



Features Available in RevBayes

- There are many features available now.
- Currently we are at a beta testing stage.
- The release will be submitted after this workshop.
- RevBayes is still under

Available distributions/functions/models

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- Substitution models:

- Jukes Cantor
- Felsenstein 81
- HKY85
- General time reversible (GTR)
- Empirical Amino Acid (mtRev, ...)
- Coala
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- Rate variation among sites:

- Any mixture you want (e.g., gamma)!

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- Clock models:
 - strict clock
 - iid clock rates (e.g., independent gamma rates)
 - mixture distributions (e.g., UCLN and UCE)
 - autocorrelated lognormal
 - RLC
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- Inference:

- Metropolis-Hastings (MCMC and reversible jump MCMC)
- Metropolis-coupled MCMC
- Power-posteriors (Path-sampling and stepping-stone-sampling)

Performance Study: Primates

Primates:

- 12 taxa
- 898 sites
- 412 patterns

MCMC:

- burnin of 10^5
- chain length of 10^6
- only substitution model parameters are updated

	HKY	HKY+G	GTR	GTR+G
BEAST v1.8	95.8	325.5	110.3	354.9

*** MrBayes used two runs because the single run does not allow to set tree proposals to 0.

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RevBayes (char specific)	92.6	269.8	120.8	326.4
RevBayes (SSE double precision)	65.1	246.7	114.8	302.6

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Performance Study: MCMC Shortcuts

Primates:

- 12 taxa
- 898 sites
- 412 patterns

Cetaceans:

- 71 taxa
- 1140 sites
- 578 patterns

MCMC:

- burnin of 10^5
- chain length of 10^6
- only topology or node ages are updated

	Narrow	NodeSlide	Narrow	NodeSlide
BEAST v1.8	3:19	3:41	6:49	10:40
RevBayes (SSE double precision)	1:29	1:38	4:18	5:29

Primates

Cetaceans

A brief intro to Rev

- Rev - The computing language used within RevBayes:
 - is an interactive environment
 - basic syntax is inspired by 'R' (and partially by BUGS)
 - aimed to built graphical models
 - provides standard 'easy-to-use' math-functions

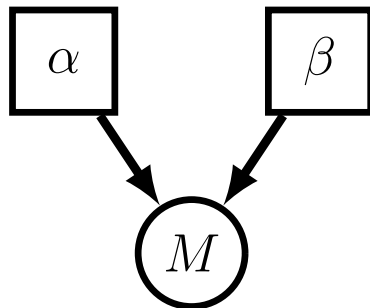

```
observations <- [<your data go here>]
```

α β

```
observations <- [<your data go here>]
```

```
alpha <- 3.0
```

```
beta <- 1.0
```

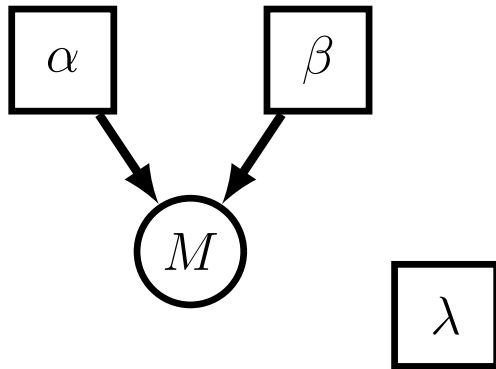


```
observations <- [<your data go here>]
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beta <- 1.0
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```
M ~ dnGamma(alpha, beta)
```



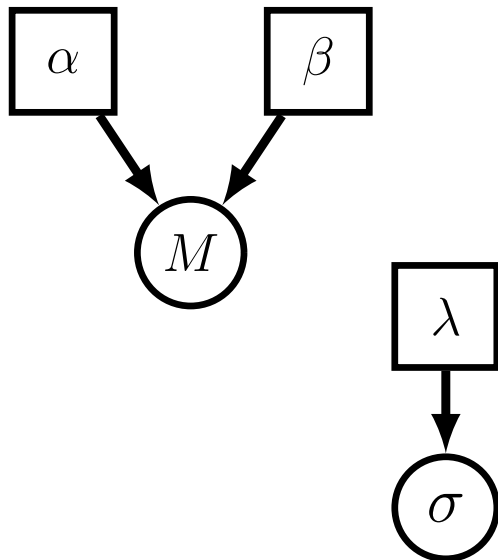
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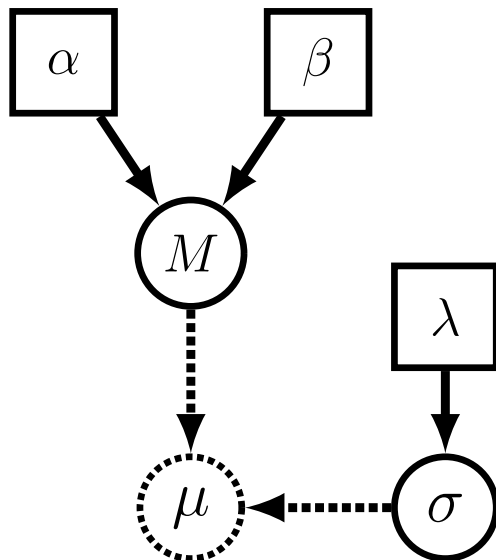
```
alpha <- 3.0
```

```
beta <- 1.0
```

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M ~ dnGamma(alpha, beta)
```

```
lambda <- 1.0
```

```
sigma ~ dnExponential(lambda)
```



```
observations <- [<your data go here>]
```

```
alpha <- 3.0
```

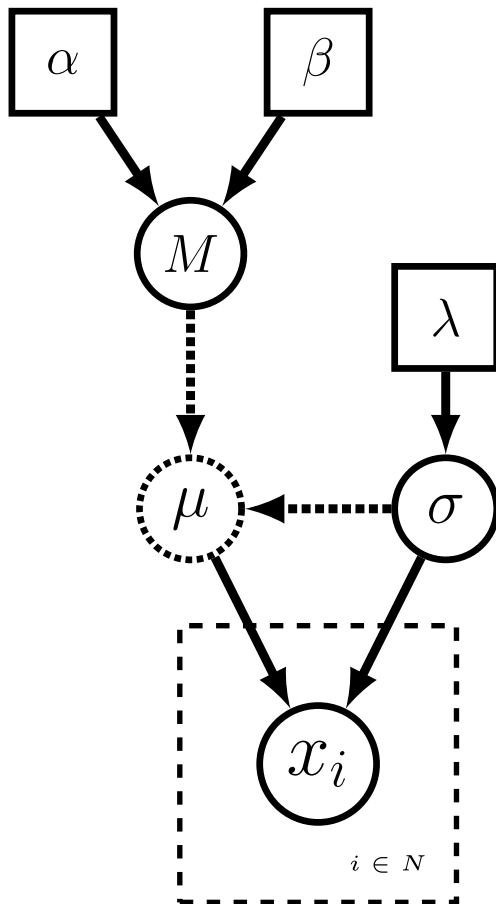
```
beta <- 1.0
```

```
M ~ dnGamma(alpha, beta)
```

```
lambda <- 1.0
```

```
sigma ~ dnExponential(lambda)
```

```
mu := ln(M) - (power(sigma, 2.0) / 2.0)
```



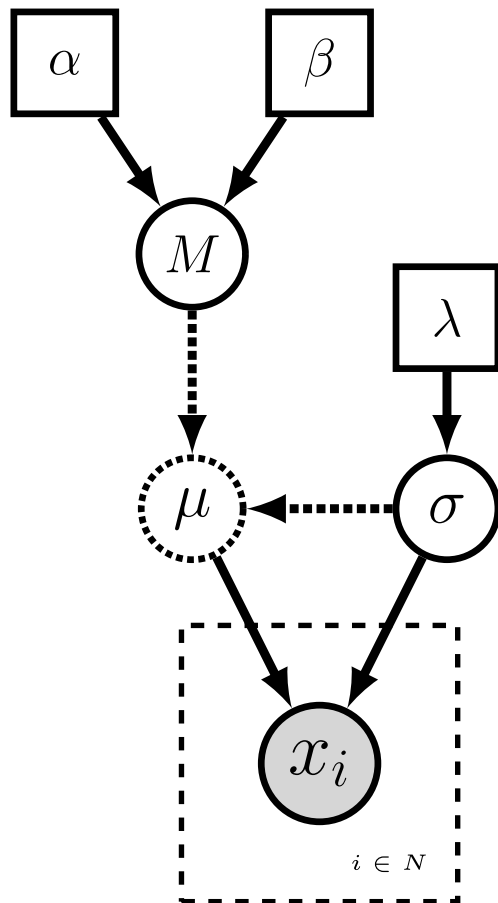
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lambda <- 1.0
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mu := ln(M) - (power(sigma, 2.0) / 2.0)

N <- observations.size()
for( i in 1:N ){
  x[i] ~ dnLnorm(mu, sigma)
}
```



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```
N <- observations.size()
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```
for( i in 1:N ){
```

```
  x[i] ~ dnLnorm(mu, sigma)
```

```
  x[i].clamp(observations[i])
```

```
}
```