

# Phylogeographic analysis of infectious disease agents based on birth–death processes

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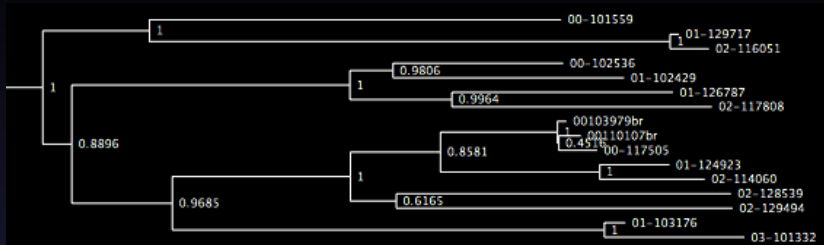
# Overview

- Motivation
- What is a "tree prior"?
- The multi-type birth–death model
- Analysis of Latvian HIV data

# Why phylogenetics?

- When and where did a virus originate in a certain host population (e.g. HIV in humans)?
- How fast does a pathogen evolve?
- What is the relationship among viral subtypes?
  
- How fast does a virus spread in a host population?
- Do intervention measures help to contain/stop an epidemic?
- How important are differences within the host population?

# Bayesian inference of time-trees



Reconstruction of evolutionary history using Bayesian Markov Chain Monte Carlo methods:

- Approximate the likelihood of the model given the data
- Sample states (model parameter combinations) from prior distribution and then sample the desired stationary distribution
- Result: Posterior distribution of the model parameters of interest (e.g. phylogenetic tree, mutation rate)

# Posterior distribution and "tree priors"

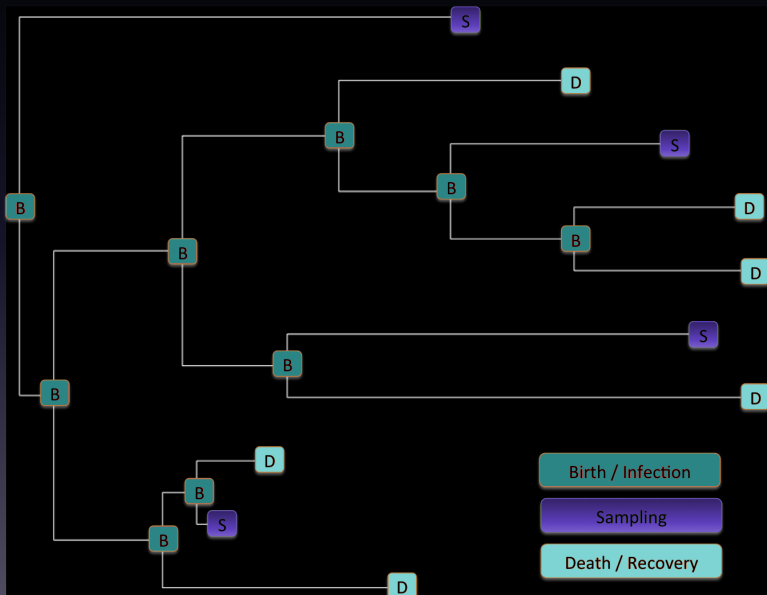
Posterior distribution of the time tree  $\mathcal{T}$ , the tree generating rates  $r$  and the model parameters  $\theta$  given the data  $\mathcal{D}$ :

$$f[\mathcal{T}, r, \theta | \mathcal{D}] \propto f[\mathcal{D} | \mathcal{T}, \theta] \underbrace{f[\mathcal{T} | r]}_{\text{"tree prior"}} f[r] f[\theta]$$

Examples of tree generating processes:

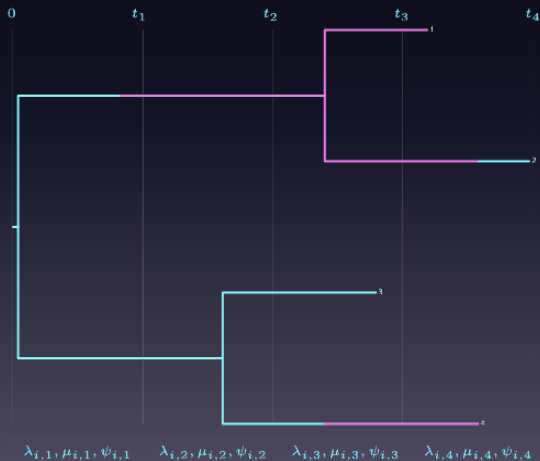
Single population	Single + rate change	Population structure
Birth–Death–Sampling model (Stadler 2010)	BDSKY (Stadler, K et al. 2013)	Multi-type Birth–death + rate change
Kingman’s coalescent (Kingman 1982)	BSP (Drummond et al. 2005)	Structured coalescent (Notohara 1990)

# The Birth–Death–Sampling tree prior



# A tree prior for 'coloured' trees

Birth–death–sampling process  $\mathcal{N} = [\mathcal{N}_1, \dots, \mathcal{N}_m]$ , where  $\mathcal{N}_i$  is the number of infected individuals in population  $i \in \{1, \dots, m\}$  and individuals can migrate among the discrete locations.



# Model parameters

Birth, death and sampling rates can differ between the  $n$  intervals and  $m$  states/locations:

$$\boldsymbol{\lambda} = ((\lambda_{1,1}, \dots, \lambda_{1,m}), (\lambda_{2,1}, \dots, \lambda_{2,m}), \dots, (\lambda_{n,1}, \dots, \lambda_{n,m}))$$

$$\boldsymbol{\mu} = ((\mu_{1,1}, \dots, \mu_{1,m}), (\mu_{2,1}, \dots, \mu_{2,m}), \dots, (\mu_{n,1}, \dots, \mu_{n,m}))$$

$$\boldsymbol{\psi} = ((\psi_{1,1}, \dots, \psi_{1,m}), (\psi_{2,1}, \dots, \psi_{2,m}), \dots, (\psi_{n,1}, \dots, \psi_{n,m}))$$

Reparametrization into *reproduction ratio*  $R$ , *become non-infectious rate*  $\delta$  and *sampling proportion*  $s$

$$R_{i,j} = \frac{\lambda_{i,j}}{\mu_{i,j} + \psi_{i,j}}, \quad \delta_{i,j} = \mu_{i,j} + \psi_{i,j}, \quad s_{i,j} = \frac{\psi_{i,j}}{\delta_{i,j}}$$

Migration matrix  $\boldsymbol{M}$  is constant over time:

$$\boldsymbol{M} = \begin{bmatrix} m_{11} & m_{12} & \cdots & m_{1n} \\ m_{21} & m_{22} & \cdots & m_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ m_{n1} & m_{n2} & \cdots & m_{nn} \end{bmatrix}$$



# The likelihood of a 'coloured' tree

The probability density  $f$  of a tree is

$$f(\mathcal{T}|\boldsymbol{\lambda}, \boldsymbol{\mu}, \boldsymbol{\psi}, \mathbf{M}, t) = \sum_{k=1}^m s_k \cdot g_{l(t),k}^e(t)$$

with  $l(t) = i$  iff  $t_{i-1} \leq t < t_i$  and  $s_k$  denoting the probability that an individual is in state  $k$ , and the probability density  $g_{i,k}^e(t)$  at time  $t_{k-1} \leq t < t_k$  that an individual in state  $i \in \{1, \dots, n\}$  evolved as observed in the tree.

# Application to Latvian HIV data

## Data set:

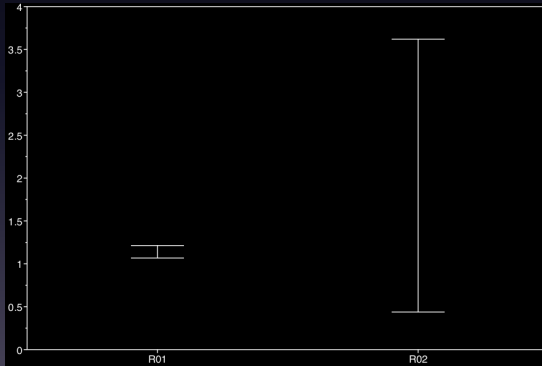
- 130 sequences of HIV subtype A from Latvia
- Matrix protein p17
- 65 sequences from HET's, 65 from IDU's
- Subsample of previously published data by Balode et al. 2004, 2012

## Questions:

- What are the epidemiological characteristics of both risk groups?
- Is the IDU sub epidemic 'stronger' than the HET subepidemic?
- Does the HET subepidemic depend on IDU's?  
cf. Stadler, Bonhoeffer 2013

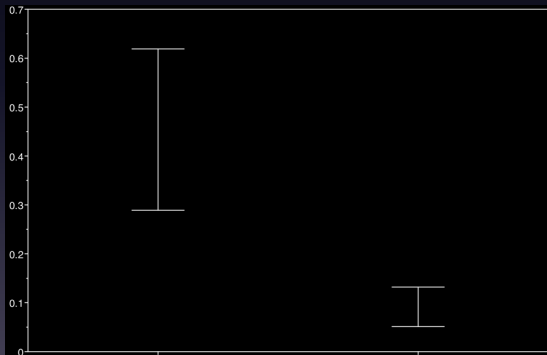
# Preliminary results - $R$

	$R_{HET}$ (median)	$R_{IDU}$ (median)
MCMC	1.14	1.63
ML*	0.38	1.13



# Preliminary results - migration rates

HET  $\rightarrow$  IDU    IDU  $\rightarrow$  HET  
0.43                0.086



# Summary

- Multi-type birth–death model provides a tree prior for phylogeographic analysis, including migration rate estimation
- Enables simulation of 'coloured' phylogenies
- Epidemiological rates can differ among locations
- Simultaneous reconstruction of the phylogeny and epidemiological rates
- Latvian HIV requires further analysis:
  - Use exactly the same data set as Stadler, Bonhoeffer 2013
  - Allow infections between compartments instead of migration
  - Compare with V3 region
  - Investigate large become non-infectious rate (med  $\delta = 2.54$ )

# Thank you for your attention

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