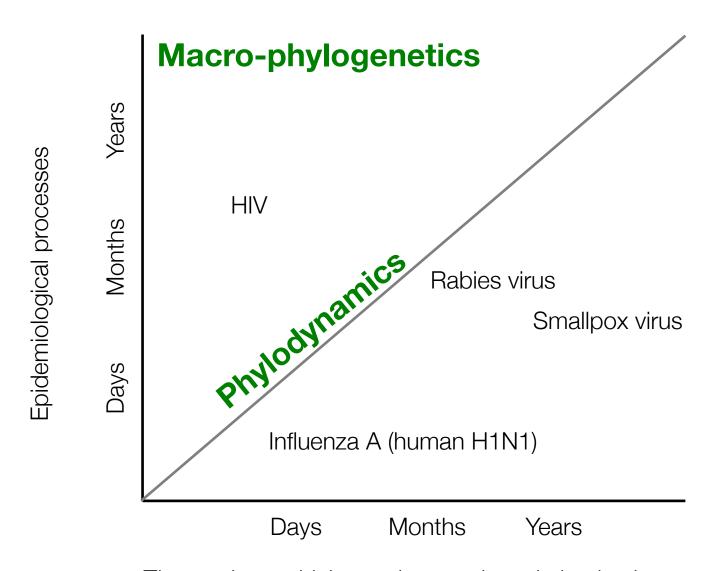
Lecture 2.4: Infectious disease phylodynamics

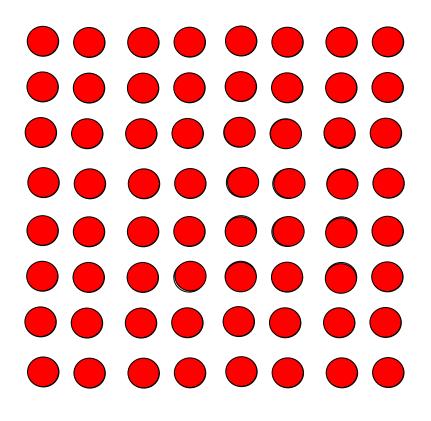


Timescale at which novel genomic variation is observed

From Biek et al. 2013 Trends Ecol Evol

The basic reproductive number

The basic reproductive number: R_0



number of secondary

susceptible population

infections in a **fully**

 R_0 is the average

 $R_0>1$ Infection can spread

 R_0 <1 Infection will die out

 R_0 for some diseases

Measles >12

HIV 2-5

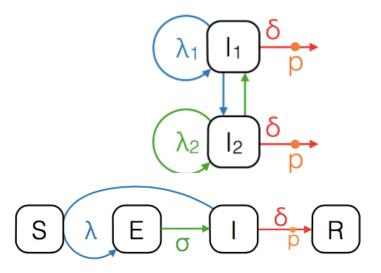
Ebola 2.2 (Stadler et al. 2014)

TB 3.4 (Tanaka et al. 2006)

The basic reproductive number: R_0

- Can depend on many factors:
 - Duration of infection
 - Incubation period
 - Host immunity
 - Behavioral changes

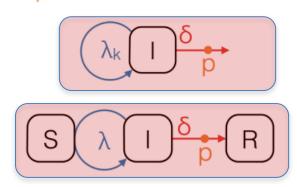
Some of these can be modeled using compartmental models:



λ — infection rate

 \(\begin{align*}
 \limits_{\text{order}} & \text{becoming-noninfectious rate}
 \)

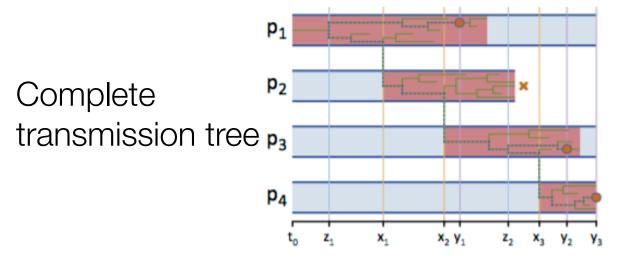
p — sampling probability



From Stadler et al. 2014 PLOS Current Outbreaks

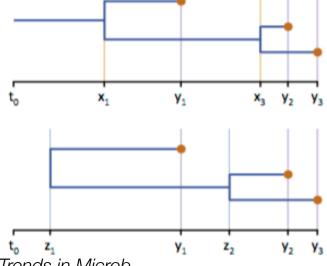
Phylogenetic epidemiology

Phylogenetics trees and infection



Sampled transmission tree

Reconstructed genealogy



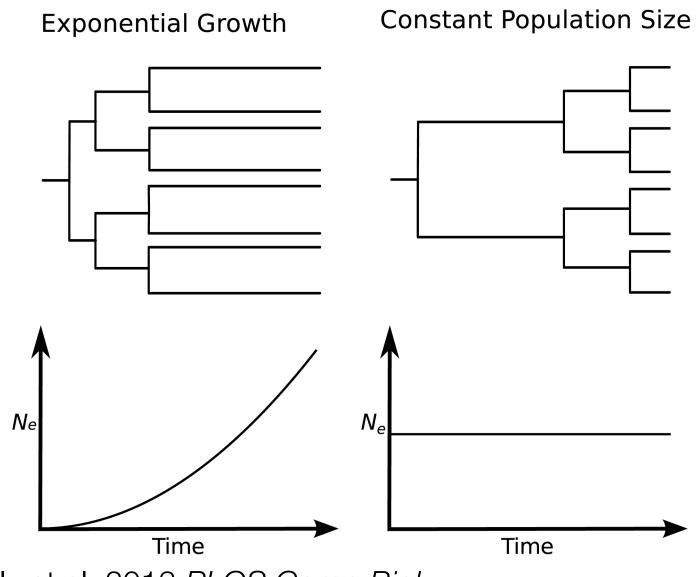
Account for sampling fraction

Assume that the sampled tree and reconstructed genealogy are equal

The genealogies have no information about who infected whom

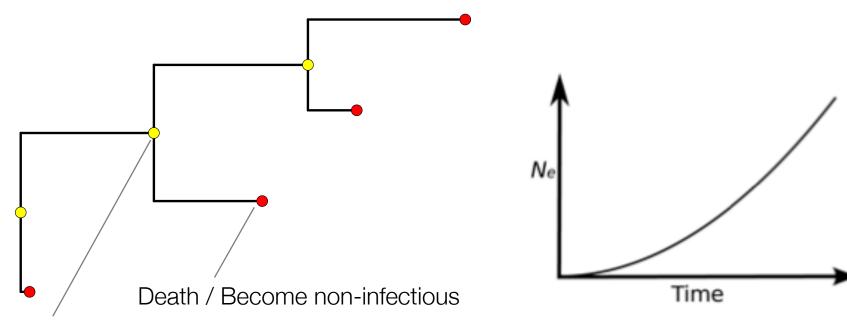
From du Pleiss and Stadler 2015 Trends in Microb

Estimating population size: the coalescent



From Volz et al. 2013 PLOS Comp Biol

Estimating R_0 : coalescent phylodynamics



Transmission event

$$N(t) = N_0 e^{-rt}$$

r: exponential growth rate N_0 : initial population size N(t): population size at time t

$$r = (R_0 - 1) / D$$

 R_0 : basic reproductive ratio

D: duration of infection

Estimating R0 using the exponential coalescent model in BEAST

r: growth rate

Φ: scaled population size

 λ : birth rate (transmission)

 δ : become uninfectious rate $\rightarrow 1/\delta$: duration of infection

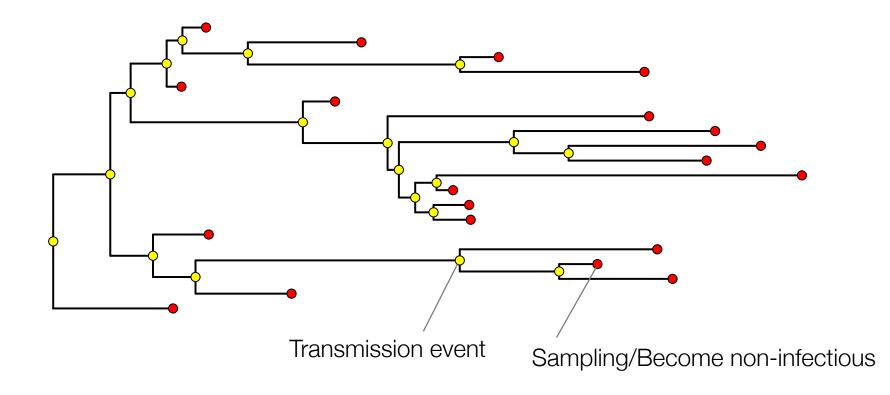
I(0): Number of infected individuals at present

$$r = \lambda - \delta$$

$$\Phi = I(0) / (2*\lambda)$$

$$R_0 = \lambda / \delta$$

Estimating R_0 : Birth-Death models



 λ : transmission rate (birth)

$$R_0 = \lambda / \delta$$

 δ = sampling + death/recovered

Estimating R0 using the birth-death model in BEAST

 λ : birth rate (transmission)

 δ : become uninfectious rate $\rightarrow 1/\delta$: duration of infection

 Ψ : sampling rate

 μ : death rate

p: sampling proportion = $\Psi / (\Psi + \mu)$

$$\delta = \Psi + \mu$$
$$R_0 = \lambda / \delta$$

Likelihood depends on $\lambda^*\delta^*p$ and λ - δ in both models (BD or CE)

Skyline methods

- Coalescent and Birth-Death methods assume constant parameters
- This assumption can be relaxed
 - Birth Death Skyline
 - Coalescent Skyline

 $R_{\rm e}(t)$

 $R_e(0) = R_0$

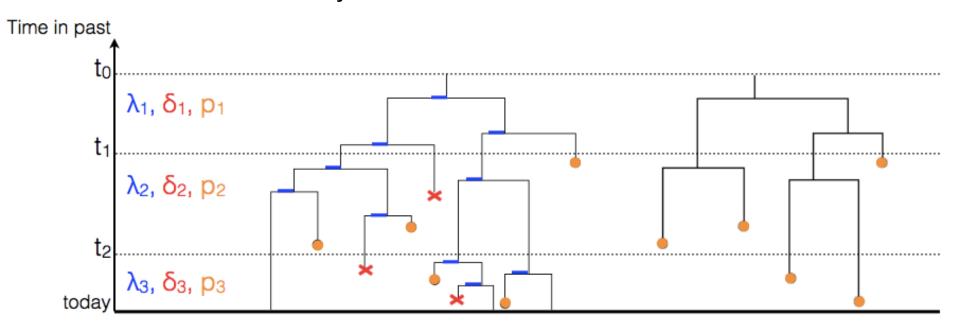
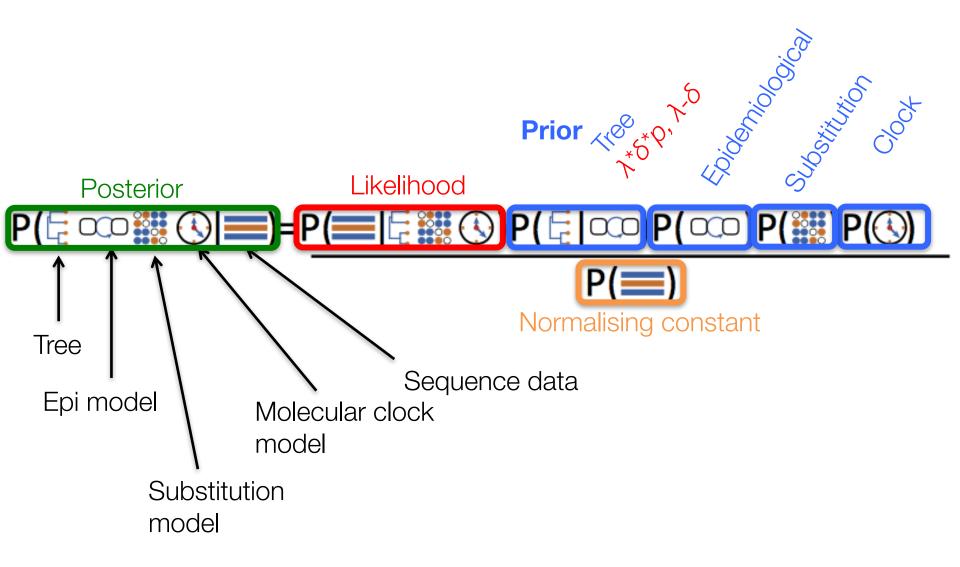


Image modified from Tanja Stadler's slides

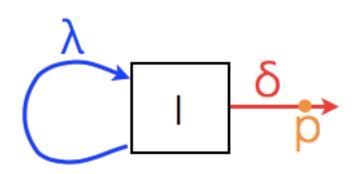
Modelling framework

Bayesian approach



Compartmental models

Birth-Death



$$R_0 = \lambda / (\delta)$$

 $\lambda = transmission rate$

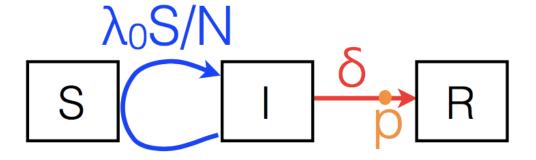
 δ = becoming non-infectious rate

 $\rho = sampling$

From Stadler et al. 2014 PLOS Current Outbreaks

Compartmental models

Birth-Death Susceptible-infected-recovered

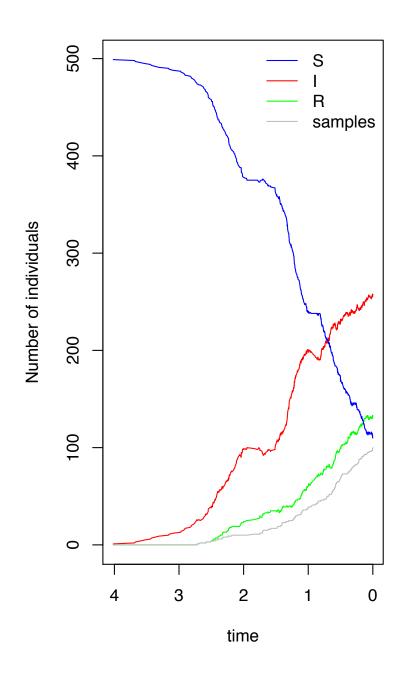


 $\lambda = transmission rate$

 δ = becoming non-infectious rate

 $\rho = sampling$

From Stadler et al. 2014 PLOS Current Outbreaks



Transmission rate (birth): β

Become uninfectious rate: γ

Susceptible pop. size: n_s

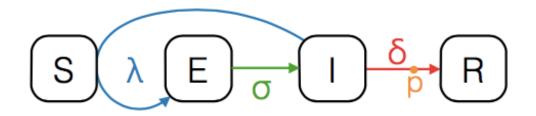
$$R_e = \beta^* n_s / \gamma$$

$$R_0 = \beta^* n_s(0) / \gamma$$

SIR and BD are equivalent when $S \rightarrow N$

Compartmental models

Birth-Death Exposed-infected-recovered



 $\lambda = transmission rate$

 δ = becoming non-infectious rate

ρ = sampling probability

 σ = incubation rate

From Stadler et al. 2014 PLOS Current Outbreaks

Compartmental models

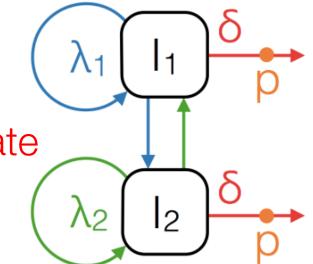
Birth-Death super-spreading

 λ_1 = transmission rate 1

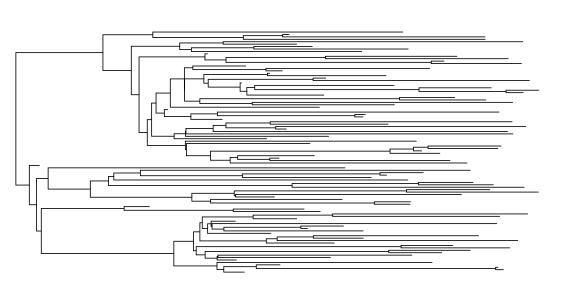
 λ_2 = transmission rate 2

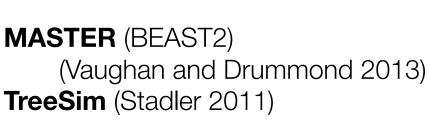
 δ = becoming non-infectious rate

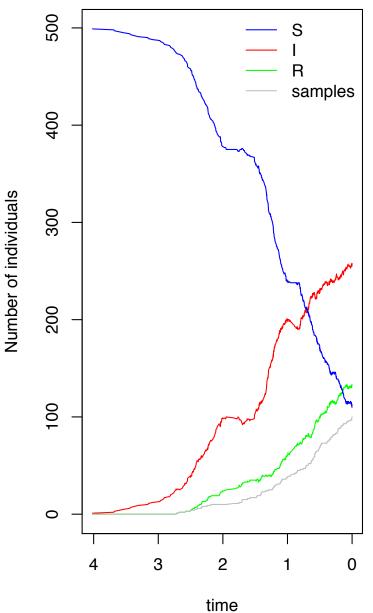
 ρ = sampling probability



Simulating trajectories







Useful references

Kühnert, Denise, et al. "Simultaneous reconstruction of evolutionary history and epidemiological dynamics from viral sequences with the birth–death SIR model." *Journal of the Royal Society Interface* 11.94 (2014): 20131106.

du Plessis, Louis, and Tanja Stadler. "Getting to the root of epidemic spread with phylodynamic analysis of genomic data." *Trends in microbiology* 23.7 (2015): 383-386.

Boskova, Veronika, Sebastian Bonhoeffer, and Tanja Stadler. "Inference of epidemiological dynamics based on simulated phylogenies using birth-death and coalescent models." PLoS computational biology 10, no. 11 (2014): e1003913.

Go to Practical: Phylogenetic epidemiology in BEAST