Phylodynamics: phylogenetics meets epidemiological modeling

Molecular Epidemiology of Infectious Diseases
Lecture 10

April 4th, 2022

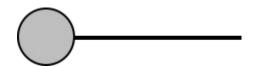
The road here

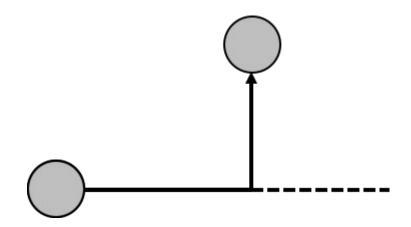
Two weeks ago: Modeling epidemic dynamics with SIR models

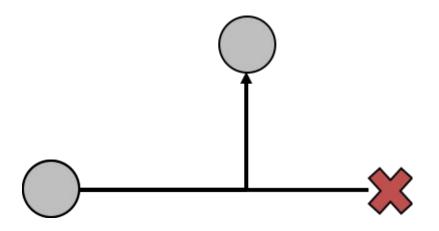
Last week: Stochastic models for simulation and inference

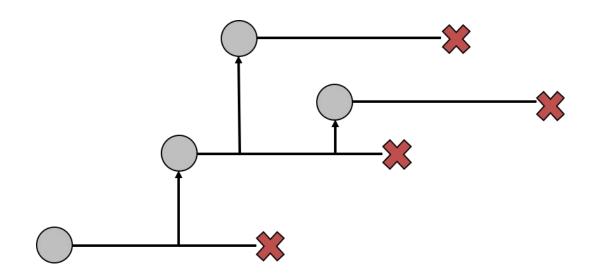
This week: putting everything together with phylodynamic modeling

Now we get to put all the pieces back together again!



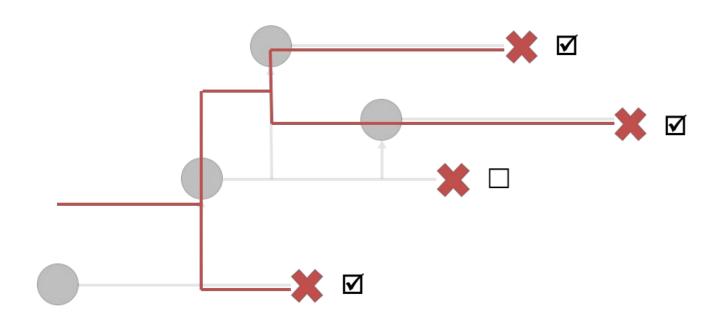


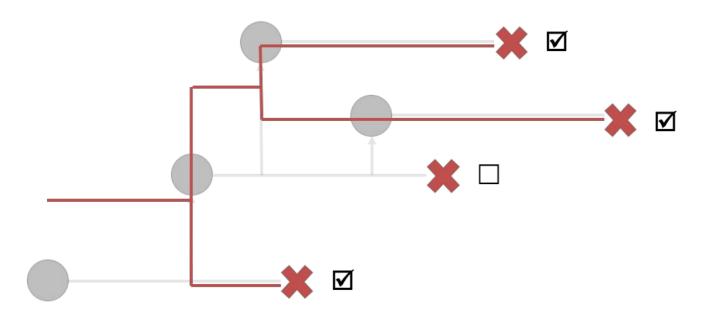




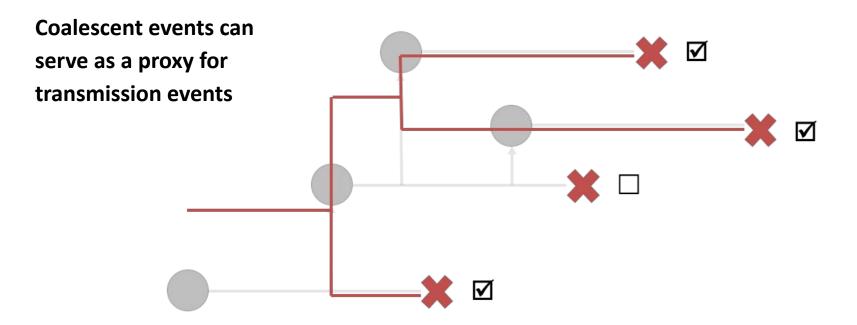
Transmission tree





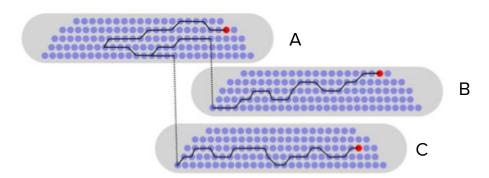


Pathogen phylogenies recapitulate major features of the underlying transmission tree



Within-host diversity

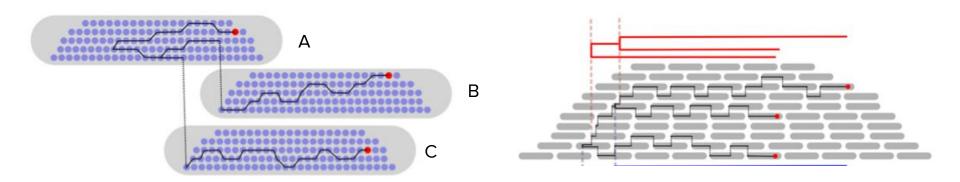
The branching structure of the phylogeny will depend on the timing and order of coalescent events within hosts



Ympa et al. (Genetics, 2013)

The macroscopic view

If we assume a single pathogen lineage resides in each infected hosts, coalescent events will closely coincide with transmission events.

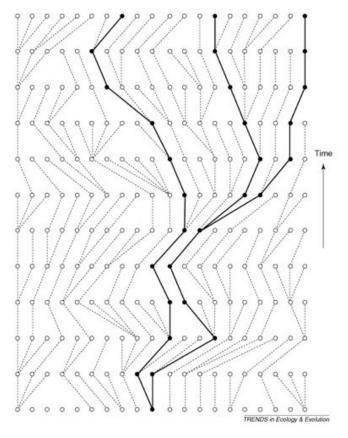


Ympa et al. (Genetics, 2013)

Basic coalescent theory

The probability of two lineages coalescing per generation is:

$$p_{coal} = \frac{1}{N}$$



Kuhner et al. (2008)

Basic coalescent theory

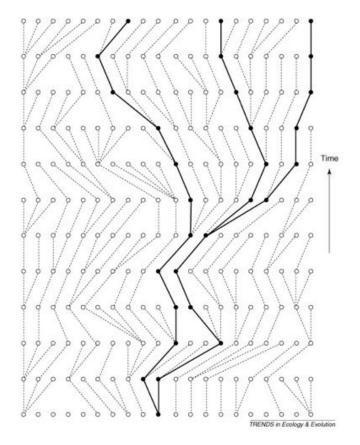
The probability of two lineages coalescing per generation is:

$$p_{coal} = \frac{1}{N}$$

The rate of coalescence in continuous time:

$$\lambda = p_{coal} = \frac{1}{N}$$

$$Pr(X = t) = \lambda e^{-\lambda t}$$

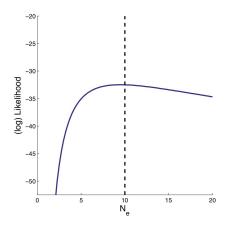


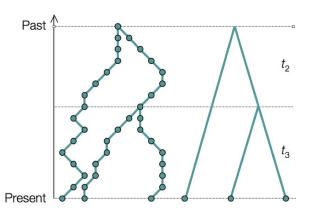
Kuhner et al. (2008)

Coalescent-based inference

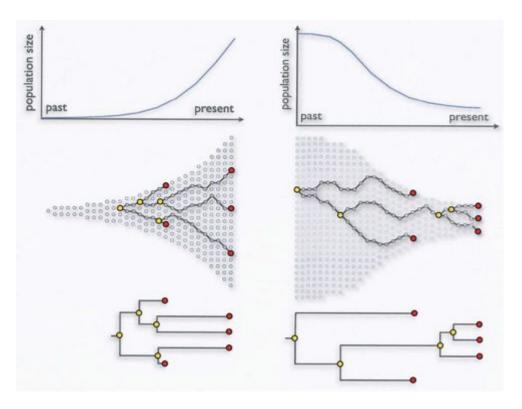
We can therefore infer demographic parameters like $N_{_{\!P}}$ from a known phylogeny.

$$L(T|N_e) = \frac{1}{N_e^{(n-1)}} \prod_{k=2}^{n} \exp\left(-\frac{\binom{k}{2}}{N_e} t_k\right)$$





The signal of population size change

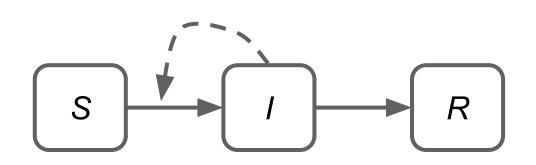


But are these appropriate coalescent models for an infectious pathogen?

The SIR model

Under the SIR model the transmission or "birth rate" of new pathogen lineages depends on the incidence which depends on both S and I.

$$\begin{array}{rcl} \frac{dS}{dt} & = & -\beta SI \\ \frac{dI}{dt} & = & \beta SI - \gamma I \\ \frac{dR}{dt} & = & \gamma I \end{array}$$



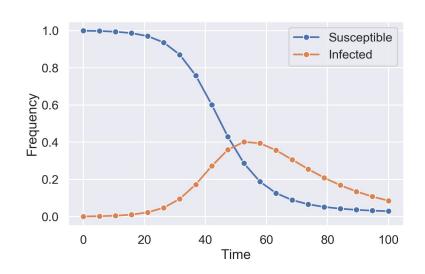
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The SIR coalescent model

Starting from first principles, we can intuitively reason that **two things must occur** in order for a pair of lineages (currently residing in two different infected hosts) to coalesce:

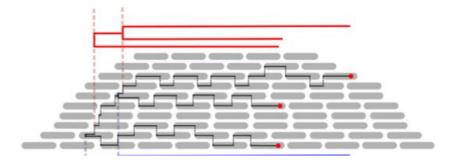
1. A transmission event must occur somewhere in the host population.

Our pair of lineages must reside in the two infected hosts resulting from the transmission event.

The SIR coalescent model

Starting from first principles, we can intuitively reason that **two things must occur** in order for a pair of lineages (currently residing in two different infected hosts) to coalesce:

1. A transmission event must occur somewhere in the host population.



SIR-type coalescent model

The total rate at which transmission/coalescent events occur in the population depends on the incidence of new infections:

$$f(t) = \beta S(t)I(t)$$

The SIR coalescent model

Starting from first principles, we can intuitively reason that **two things must occur** in order for a pair of lineages (currently residing in two different infected hosts) to coalesce:

2. Our pair of lineages must reside in the two infected hosts resulting from the transmission event.

$$S+I \xrightarrow{\beta} 2I$$

Probability of two lineages coalescing

What is the probability that one pair of lineages coalesces rather than any other pair of lineages?

There are I(t) total pathogen lineages in the host population at time t.

The total number of pairs that could possibly coalesce is:

Total pairs =
$$\binom{I(t)}{2} = \frac{I(t)(I(t)-1)}{2}$$

The probability that our pair of lineages coalesces is therefore:

$$p_{coal} = \frac{1}{\text{Total pairs}} = \frac{2}{I(t)(I(t) - 1)} \approx \frac{2}{I(t)^2}$$

SIR-type coalescent model

The total rate at which transmission/coalescent events occur in the population depends on the incidence of new infections:

$$f(t) = \beta S(t)I(t)$$

Given a transmission event occurs, the probability that two particular lineages coalesce is:

$$p_{coal} \approx \frac{2}{I(t)^2}$$

The pairwise coalescent rate is therefore:

$$\lambda(t) = \frac{2\beta S(t)I(t)}{I(t)^2} = \frac{2\beta S(t)}{I(t)}$$

Volz et al. (Genetics, 2009) Volz (Genetics, 2012)

Comments on the SIR coalescent model

The rate at which two pathogen lineages coalesce not only depends on the infected population size but also the transmission rate!

This means that the coalescent rate will peak when incidence is high but prevalence is low.

What makes these models powerful is that we can now fit epidemiological models directly to pathogen phylogenies.

Host population structure

Host populations are almost always structured into different subpopulations:

- Age structure
- Gender structure
- Risk/contact structure
- Spatial structure
- Stages of disease progression

The problem with population structure

Standard coalescent models assume that all lineages in the tree are **exchangeable**.

Exchangeability here means that any lineage is equally likely to coalesce with any other lineage in the tree.

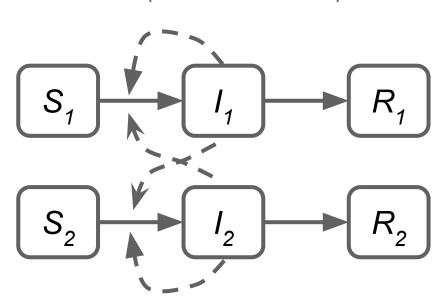
Many forms of population structure violate this key assumption.

We therefore need structured epidemiological/coalescent models!

SIR model with multiple host classes

The SIR model can be generalized to include multiple different "compartments":

$$\frac{dI_i}{dt} = \sum_j \beta_{ji} S_i I_j - \gamma I_i$$



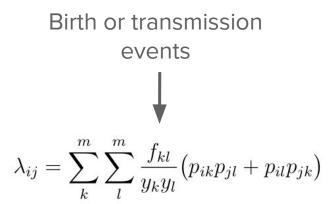
The structured coalescent model of Volz (Genetics, 2012) considers both complex epidemiological dynamics and host population structure!

Under this model, the pairwise coalescent rate for two lineages *i* and *j*:

$$\lambda_{ij} = \sum_{k=1}^{m} \sum_{l=1}^{m} \frac{f_{kl}}{y_k y_l} \left(p_{ik} p_{jl} + p_{il} p_{jk} \right)$$

$$\lambda_{ij} = \sum_{k=1}^{m} \sum_{l=1}^{m} \frac{f_{kl}}{y_k y_l} \left(p_{ik} p_{jl} + p_{il} p_{jk} \right)$$

Sum over all possible locations of both lineages



$$\lambda_{ij} = \sum_{k=1}^{m} \sum_{l=1}^{m} \frac{f_{kl}}{y_k y_l} (p_{ik} p_{jl} + p_{il} p_{jk})$$

Individuals in populations k and l

$$\lambda_{ij} = \sum_{k=1}^{m} \sum_{l=1}^{m} \frac{f_{kl}}{y_k y_l} \left(p_{ik} p_{jl} + p_{il} p_{jk} \right)$$

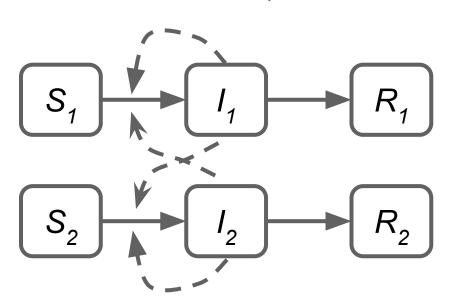


Lineage state probabilities

SIR coalescent model with structure

The Volz structured coalescent model for the SIR model with multiple host classes:

$$\lambda_{ij} = \sum_{k=1}^{m} \sum_{l=1}^{m} \frac{\beta_{kl} S_{l} I_{k}}{I_{k} I_{l}} (p_{ik} p_{jl} + p_{il} p_{jk})$$



The Volz Structured Coalescent

The structured coalescent model of Volz (Genetics, 2012) considers both complex epidemiological dynamics and host population structure!

Under this model, the pairwise coalescent rate for two lineages *i* and *j*:

$$\lambda_{ij} = \sum_{k=1}^{m} \sum_{l=1}^{m} \frac{f_{kl}}{y_k y_l} (p_{ik} p_{jl} + p_{il} p_{jk})$$

The location of each lineage back through time is **tracked probabilistically** in terms of the lineage state probabilities $p_{i\nu}$

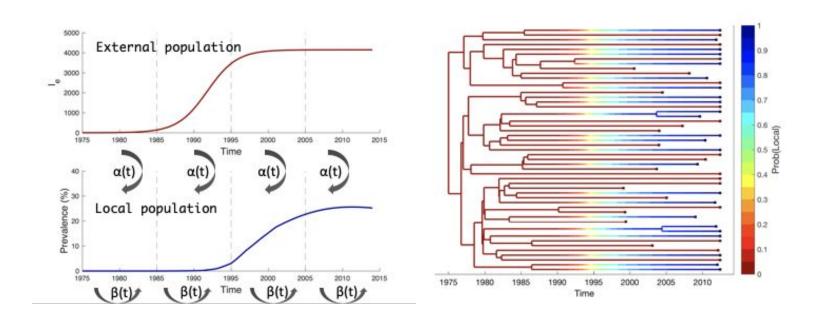
The Volz Structured Coalescent

Lineage state probabilities p_{ik} are tracked backwards in time using a system of master equations (ODEs) based on the transition rates g_{kl} :

$$\frac{d}{dt}p_{ik} = \sum_{l}^{m} \left(p_{il}g_{kl} - p_{ik}g_{lk} \right)$$

Solving for p_{ik} at any point in time gives the probability that lineage i is in state k.

Tracking lineage probabilities for a two-location SIR model



Deriving other SIR-type models

The appropriate coalescent model for nearly any type of SIR model can be easily be derived from the Volz structured coalescent.

In the general case, there are two different ways lineages can move between populations:

- Through transmission events which can result in a coalescent event
- Through migration events involving a single lineage (no coalescence).

Deriving other SIR-type models

We will use the matrix F(t) to represent the rate at which lineages move between populations due to birth or transmission events:

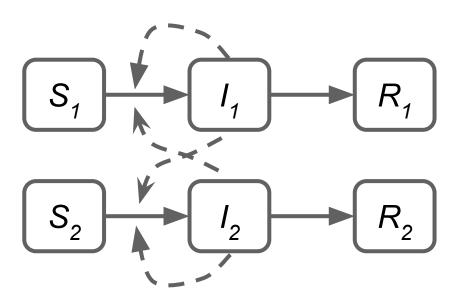
$$oldsymbol{F(t)} = egin{bmatrix} f_{11} & f_{12} & \cdots & f_{1n} \ f_{21} & f_{22} & \cdots & f_{2n} \ dots & dots & \ddots & dots \ f_{n1} & f_{n2} & \cdots & f_{nn} \end{bmatrix}$$

We will use the matrix G(t) to represent the rate at which single lineages migrate between populations: $\begin{bmatrix} g_{11} & g_{12} & \cdots & g_{1n} \end{bmatrix}$

$$m{G(t)} = egin{bmatrix} g_{11} & g_{12} & \cdots & g_{1n} \ g_{21} & g_{22} & \cdots & g_{2n} \ dots & dots & \ddots & dots \ g_{n1} & g_{n2} & \cdots & g_{nn} \ \end{pmatrix}$$

Example: SIR model with two classes

$$\frac{dI_1}{dt} = \beta_{11}S_1I_1 + \beta_{21}S_1I_2 - \gamma I_1
\frac{dI_2}{dt} = \beta_{22}S_2I_2 + \beta_{12}S_2I_1 - \gamma I_2$$



The SIR matrix decomposition

If we assume infected hosts do not migrate between populations:

$$F(t) = \begin{bmatrix} \beta_{11}S_1I_1 & \beta_{12}S_2I_1 \\ \beta_{21}S_1I_2 & \beta_{22}S_2I_2 \end{bmatrix}$$

$$S_1 \qquad I_1 \qquad R_1$$

$$G(t) = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix}$$

$$S_2 \qquad I_2 \qquad R_2$$

The SIR matrix decomposition

Or if we allow infected hosts to migrate between populations at a rate *y*:

$$F(t) = \begin{bmatrix} \beta_{11} S_1 I_1 & \beta_{12} S_2 I_1 \\ \beta_{21} S_1 I_2 & \beta_{22} S_2 I_2 \end{bmatrix}$$

$$S_1 \qquad I_1$$

$$R_1$$

$$G(t) = \begin{bmatrix} 0 & \gamma_{12} \\ \gamma_{21} & 0 \end{bmatrix}$$

$$S_2 \qquad I_2$$

Example: SEIR model

The SEIR model adds an exposed compartment to the basic SIR model:

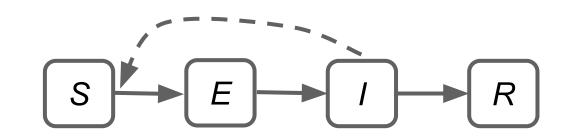
$$\frac{dS}{dt} = -\beta SI
\frac{dE}{dt} = \beta SI - \eta E
\frac{dI}{dt} = \eta E - \gamma I$$

$$\frac{dR}{dt} = \gamma I$$

The SEIR matrix decomposition

We can decompose the SEIR model into **F** and **G** rate matrices:

$$F(t) = \begin{bmatrix} 0 & 0 \\ \beta SI & 0 \end{bmatrix}$$



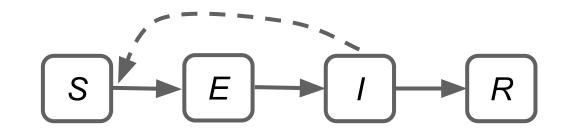
What would the G matrix look like for a SIER model?

The SEIR matrix decomposition

We can decompose the SEIR model into **F** and **G** rate matrices:

$$\boldsymbol{F(t)} = \begin{bmatrix} 0 & 0 \\ \beta SI & 0 \end{bmatrix}$$

$$\boldsymbol{G(t)} = \begin{bmatrix} 0 & \eta E \\ 0 & 0 \end{bmatrix}$$



Deriving other SIR-type models

The appropriate coalescent model for nearly any type of SIR model can be easily be derived from the Volz structured coalescent.

All we need to do is decompose the model into its component **F** and **G** matrices.

We can then compute the coalescent rates under the model:

$$\lambda_{ij} = \sum_{k=1}^{m} \sum_{l=1}^{m} \frac{f_{kl}}{y_k y_l} \left(p_{ik} p_{jl} + p_{il} p_{jk} \right)$$

And use the **G** matrix to track the movement of lineages back through time:

$$\frac{d}{dt}p_{ik} = \sum_{l}^{m} \left(p_{il}g_{kl} - p_{ik}g_{lk} \right)$$

The SEIR model in PhyDyn

PhyDyn is a BEAST2 package for fitting generic SIR models to pathogen phylogenies.

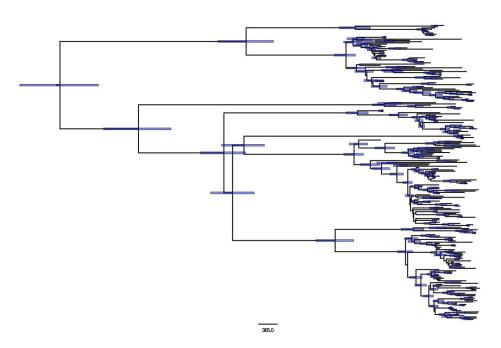
The appropriate coalescent model for a given SIR model is specified by **F** and **G** matrices .

Individual XML elements specify the rates in the **F** and **G** matrices.

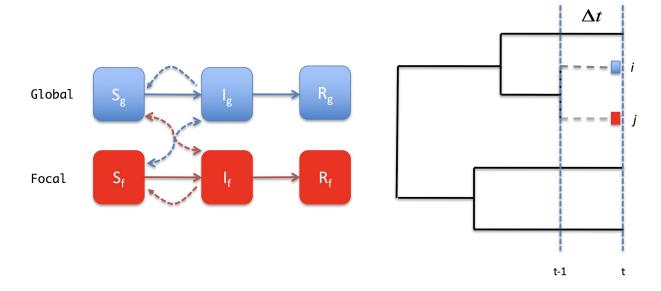
What makes these models powerful is that we can now fit epidemiological models directly to pathogen phylogenies.

Dengue in southern Vietnam





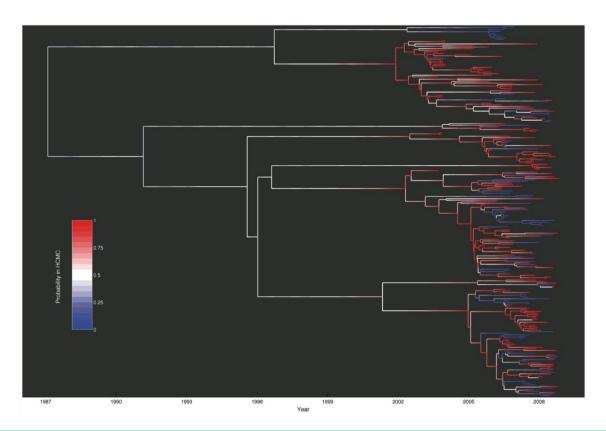
Spatial SIR model



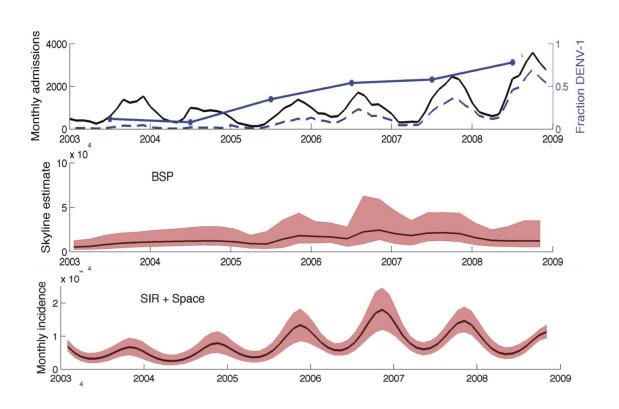
Structured coalescent model:

$$\lambda_{ij} = \sum_{k}^{m} \sum_{l}^{m} \frac{\beta_{kl} \frac{S_{l}}{N_{l}} I_{k}}{I_{k} I_{l}} \left(p_{ik} p_{jl} + p_{il} p_{jk} \right)$$

Movement of lineages



Estimates accounting for spatial structure



The PhyDyn tutorial for this week

On Wednesday, we will use the PhyDyn package to fit our own SIR-type models to pathogen phylogenies in BEAST2

First we will test our implementation of the SEI2R model in BEAST using the sequence data we simulated last week.

Then apply to the SEI2R model to some SARS-CoV-2 sequences to estimate key epidemiological parameters for Covid-19.

Does this really matter?

