Inferring transmission trees and who's infecting whom

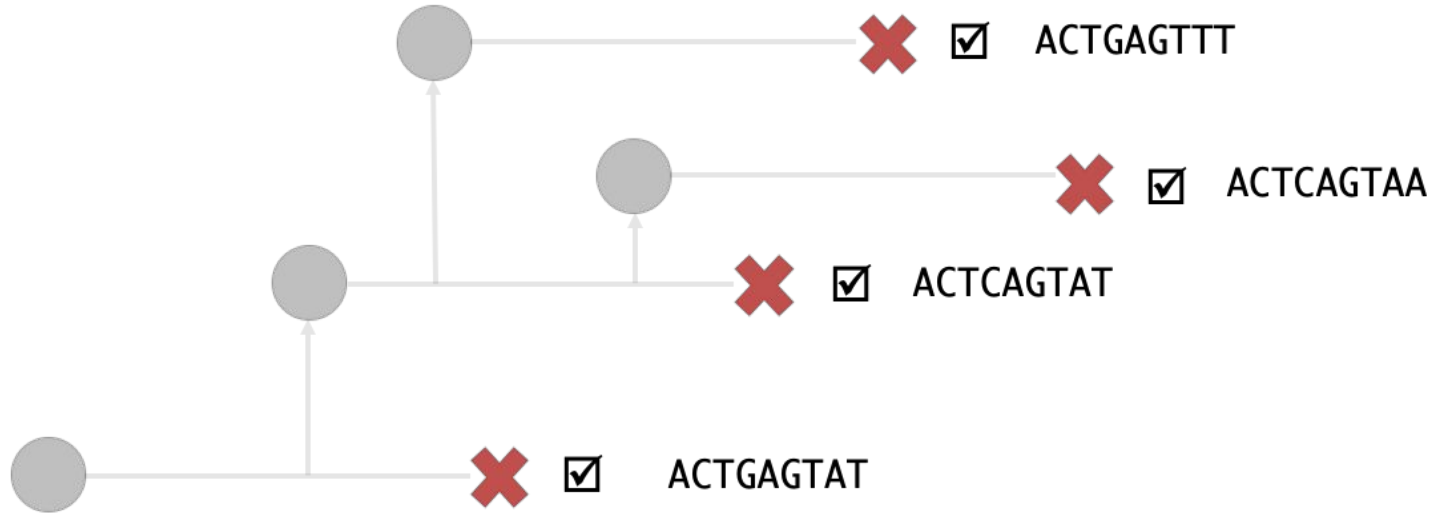Molecular Epidemiology of Infectious Diseases

Lecture 5

February 10th, 2020

So far we've been focusing on population-level transmission dynamics

Now we will turn to tracking outbreaks at the individual host-scale

A simple epidemic example

Who's infecting whom

Reconstructing who infected whom is often considered to be the "*holy grail*" of infectious disease epidemiology.

- Identifies who is actually transmitting (e.g. superspreaders)

- Identifies the characteristics of transmitters (e.g. injection drug users)

- Provides a target for control efforts and interventions

- Allows for contact-tracing to prevent further spread

Who's infecting whom

The unit of infection does not necessarily need to be individual hosts. Transmission tree methods can reconstruct spread among:

- Schools
- Villages
- Fields
- Farms

Two main approaches

1. Methods that directly estimate the underlying transmission tree

2. Methods that reconstruct pathogen phylogenies and then infer transmission routes between hosts

Two main approaches

1. Methods that directly estimate the underlying transmission tree

2. Methods that reconstruct pathogen phylogenies and then infer transmission routes between hosts
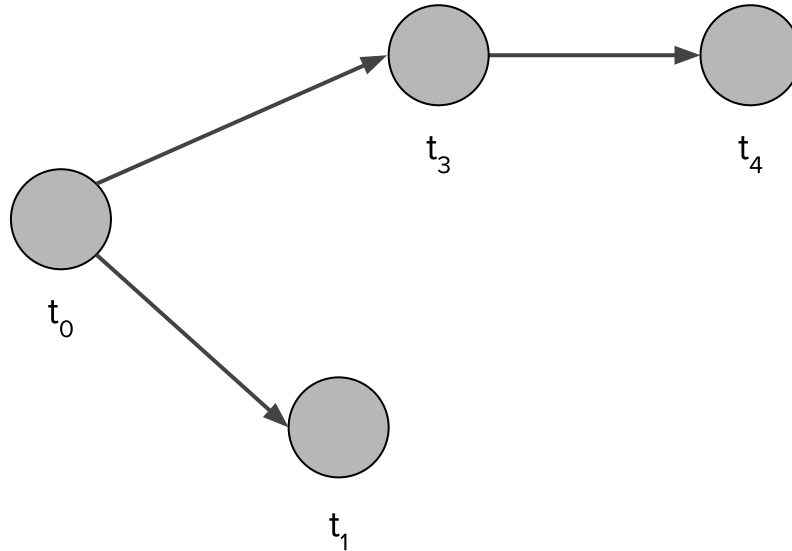
Transmission tree reconstruction

General goal is to probabilistically reconstruct likely transmission links
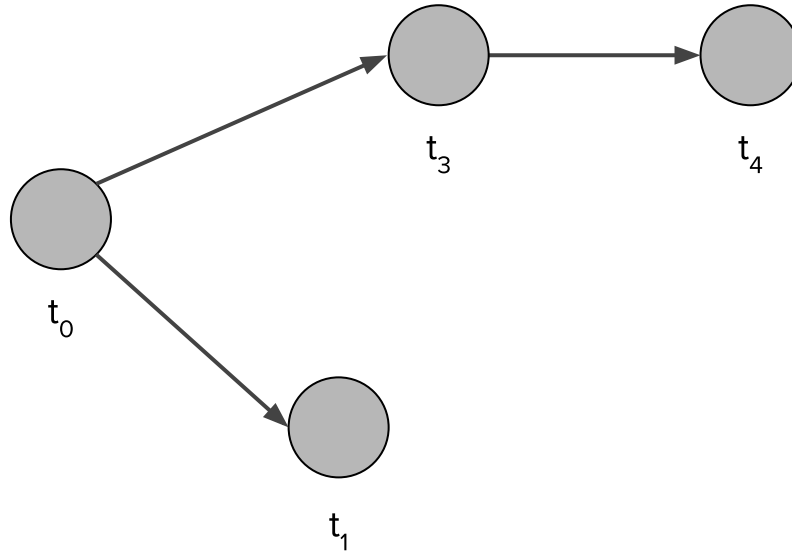
Transmission tree reconstruction

We often have data on the infection times t_1, t_2, ... t_n and sequences s_1, s_2, ... s_n sampled from each host.

Transmission tree reconstruction

We can divide the problem by thinking about the likelihood of two types of data given a proposed transmission tree:

1. The **epidemiological likelihood** of the infection times and any other spatial/temporal data we know about the infected hosts

2. The **genetic likelihood** of the sequence data

The epidemiological likelihood

The likelihood of a host infected at time t_i infecting another host at time t_j follows a generation time (serial interval) distribution:
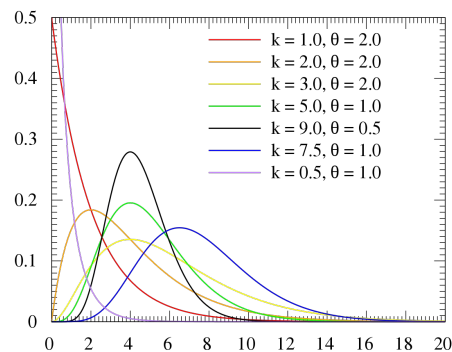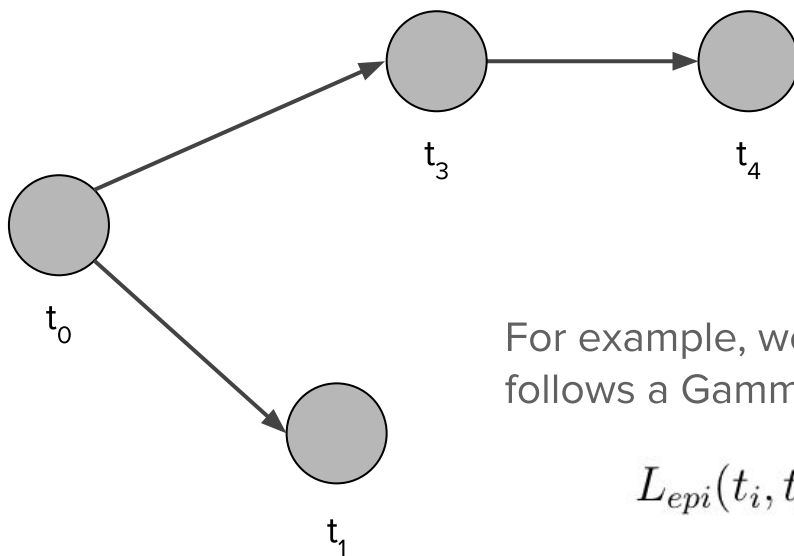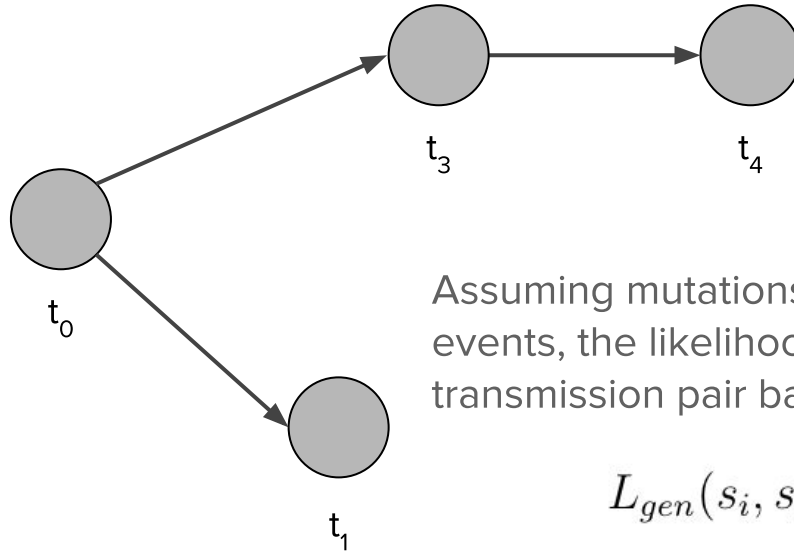

For example, we could assume the generation time follows a Gamma distribution:

$$L_{epi}(t_i, t_j) = Gamma(t_j - t_j | \alpha, \beta)$$

The genetic likelihood (simplest case)

The likelihood of sequences s_i and s_j resulting from a direct transmission between hosts i and j can be computed based on their genetic distances:

t_3

t_4

t_0

Assuming mutations happen at rate μ at transmission events, the likelihood of two sequences being in a transmission pair based on their genetic distance $d(s_i, s_j)$:

$$L_{gen}(s_i, s_j) = \mu^{d(s_i, s_j)}(1 - \mu)^{L - d(s_i, s_j)}$$

t_1

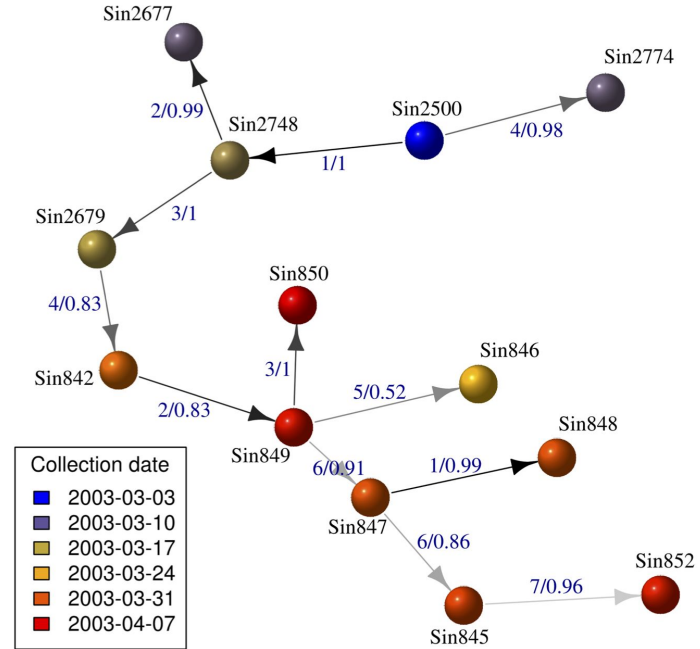Jombart *et al.* (2014)
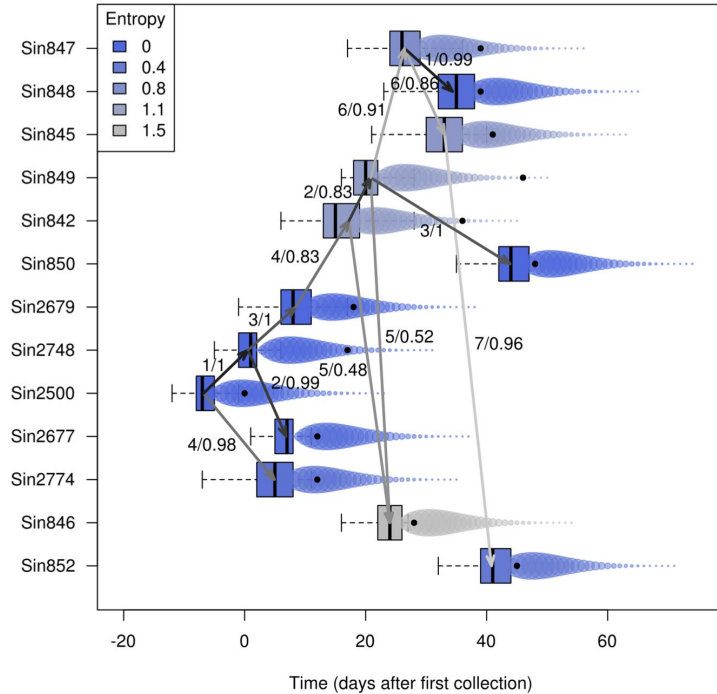
Transmission tree reconstruction

Our goal is to find the transmission tree that maximizes the **overall likelihood** of the infection times and sequence data across all transmission pairs:
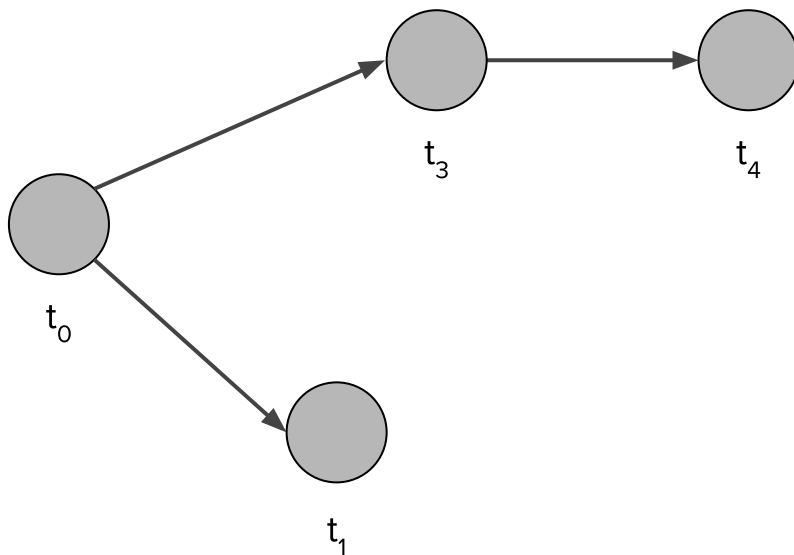
The overall likelihood can be computed as a product over all transmission pairs:

$$L(\mathcal{T}) = \prod_{i,j \in \mathcal{T}} L_{epi}(t_i, t_j) L_{gen}(s_i, s_j)$$

SARS outbreak in Singapore


Jombart *et al.* (2014)

Direct transmission tree reconstruction

Direct reconstruction generally works well when:

- Outbreaks are small and we can sample nearly all infected hosts

- Short and regular generation times

- High between-host genetic divergence but negligible within-host variation
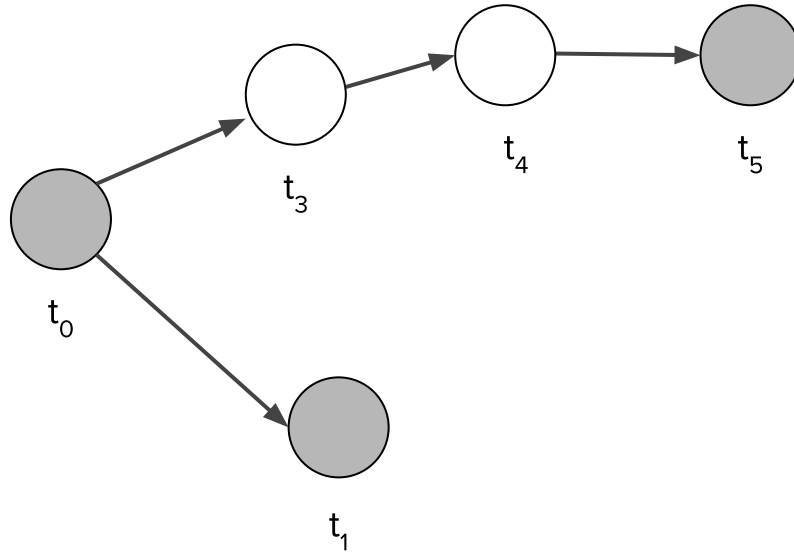
Transmission tree reconstruction

The problem is that we generally have incomplete sampling with at least some unobserved infections.
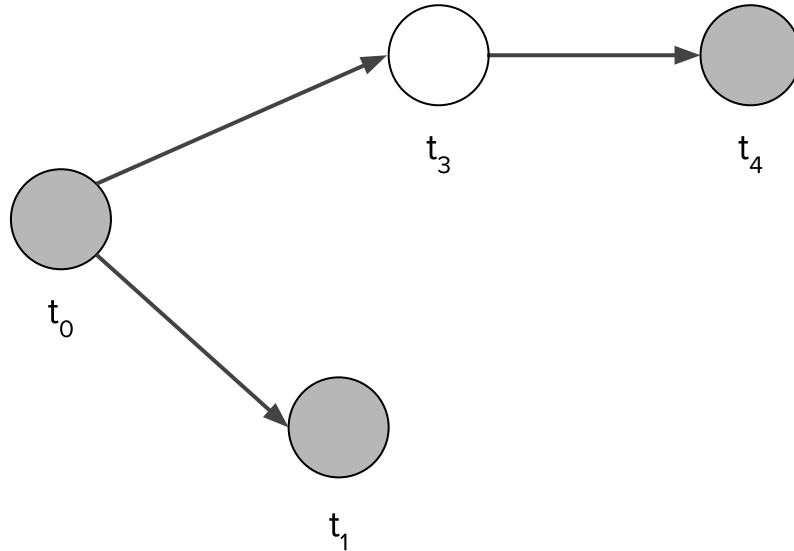
Transmission tree reconstruction

The problem is that we generally have incomplete sampling with at least some unobserved infections.
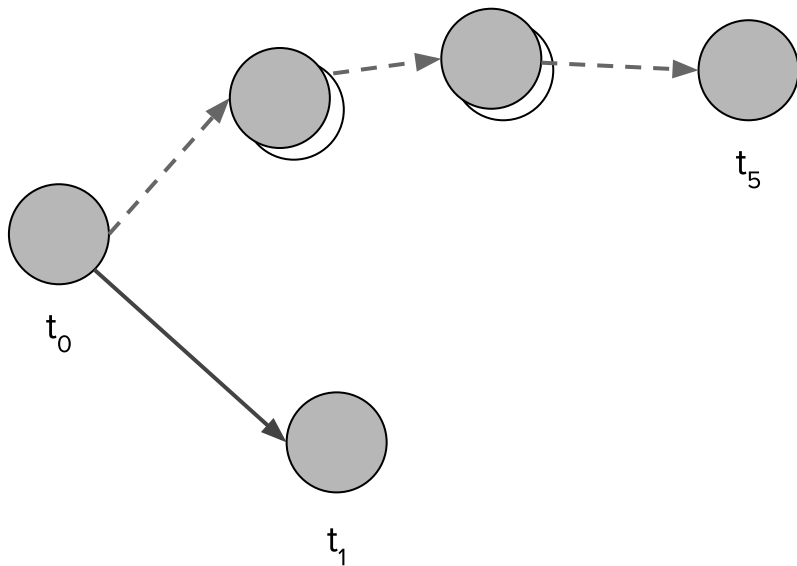
Transmission tree reconstruction

We are therefore likely to misattribute sources of infection to sampled individuals while ignoring unobserved hosts.
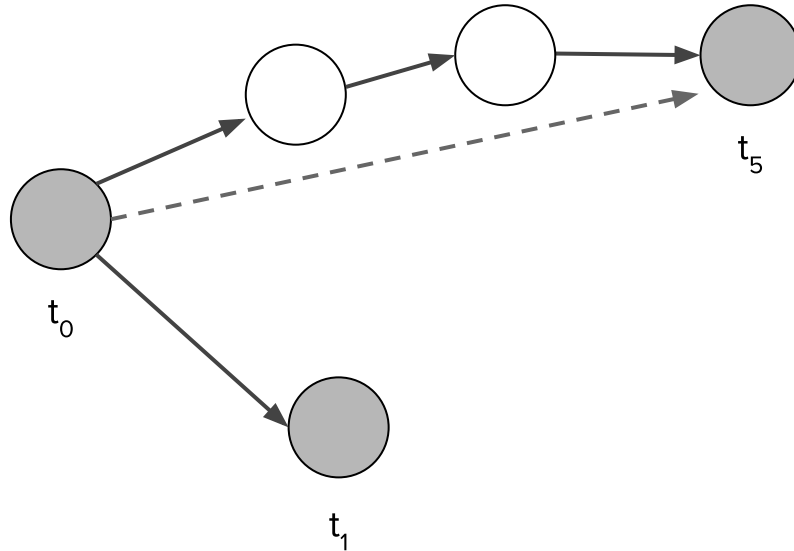
Data augmentation

We can postulate the presence of unobserved infections and impute their presence/absence and infection times as additional latent variables in the model.
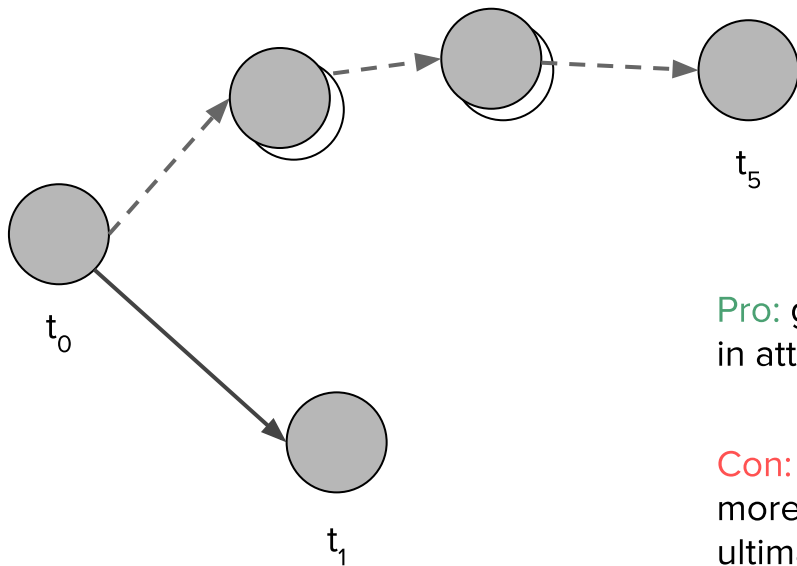
Data augmentation

We can postulate the presence of unobserved infections and impute their presence/absence and infection times as additional latent variables in the model.

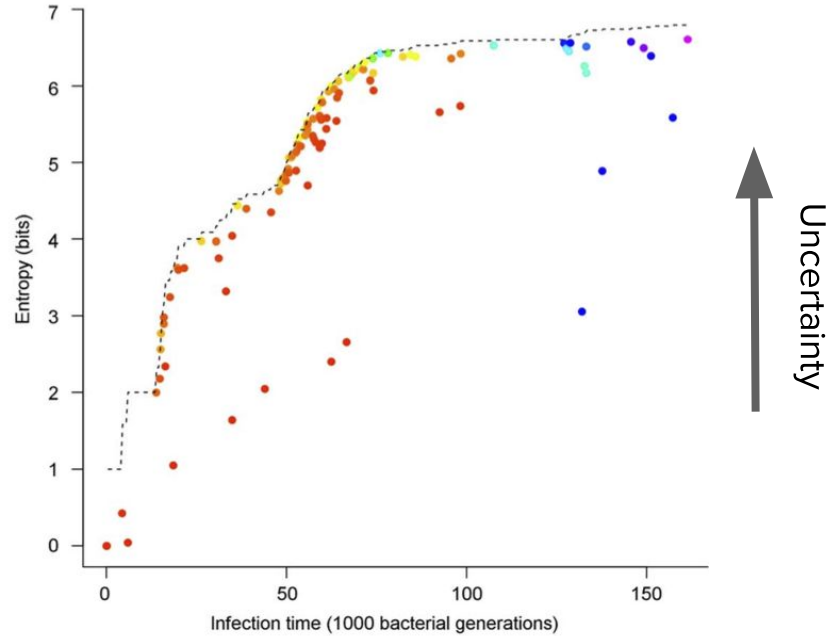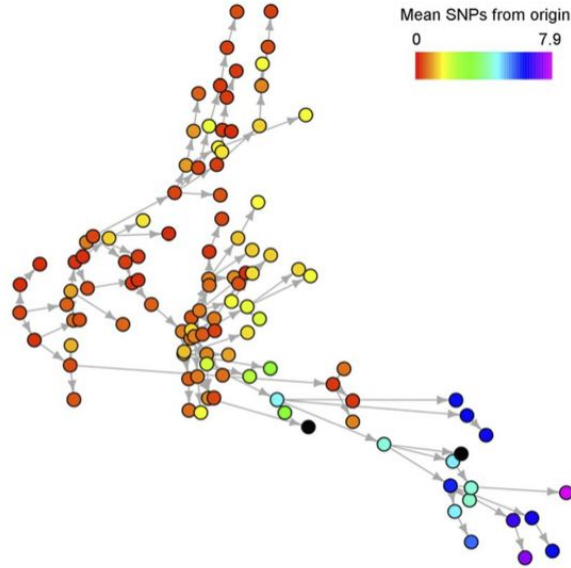
Pro: guards against overconfidence in attributing sources.

Con: uncertainty will only grow with more unsampled hosts so ultimately doomed to fail.

Direct transmission tree reconstruction

Direct reconstruction generally works well when:

- Outbreaks are small and we can sample nearly all infected hosts

- Short and regular generation times

- High between-host genetic divergence but negligible within-host variation

Effect of overlapping infections

Mean SNPs from origin
0 — 7.9

Entropy (bits)

Uncertainty

Infection time (1000 bacterial generations)

Worby *et al.* (PLoS Comp Bio, 2014)

Direct transmission tree reconstruction

General approach works well when:

- Outbreaks are small and we can sample nearly all infected hosts

- Short and regular generation times

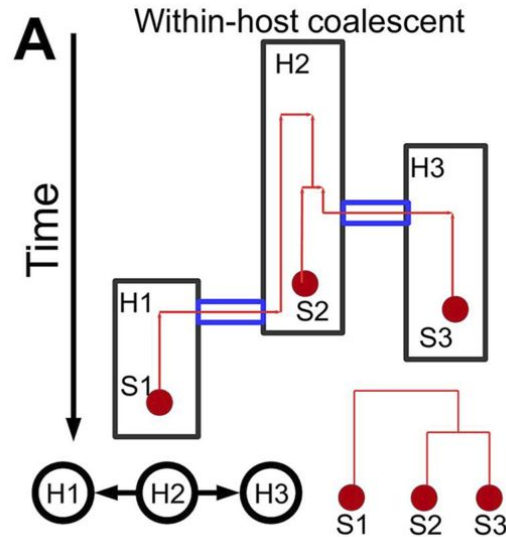- High between-host genetic divergence but negligible within-host variation

So far we have completely ignored within-host genetic diversity!
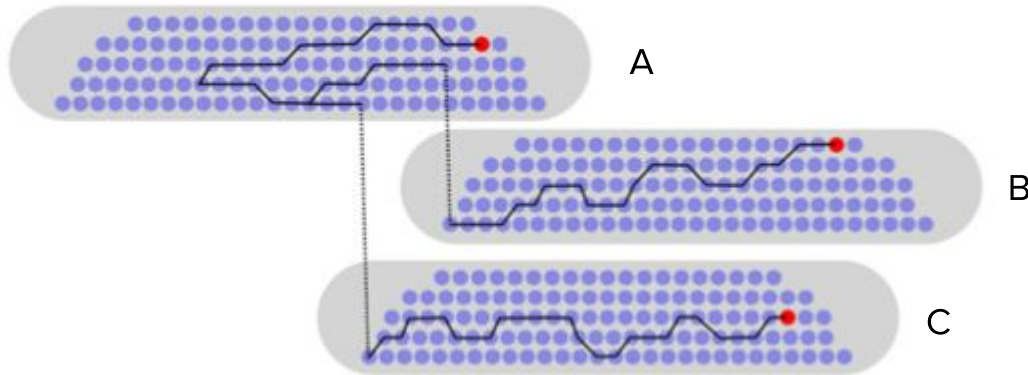
Within-host diversity

Within-host diversity can cause discordance between pathogen phylogenies and the transmission tree.

Within-host diversity

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

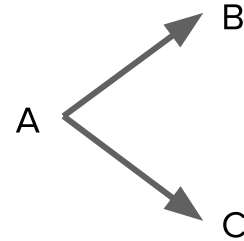
Ympa *et al.* (Genetics, 2013)

Within-host diversity

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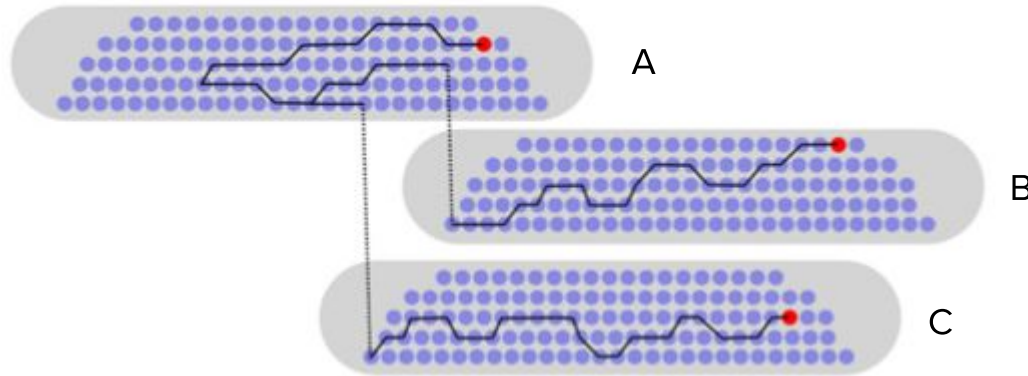
Actual transmission tree:

But all three phylogenetic trees are possible!

Ympa *et al.* (Genetics, 2013)

Within-host diversity

If two lineages coalesce at a transmission event, the coalescent event will always occur before the actual transmission event
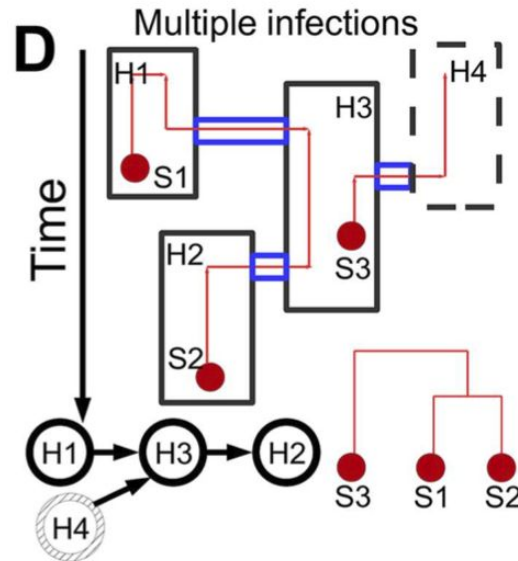

Ympa *et al.* (Genetics, 2013)
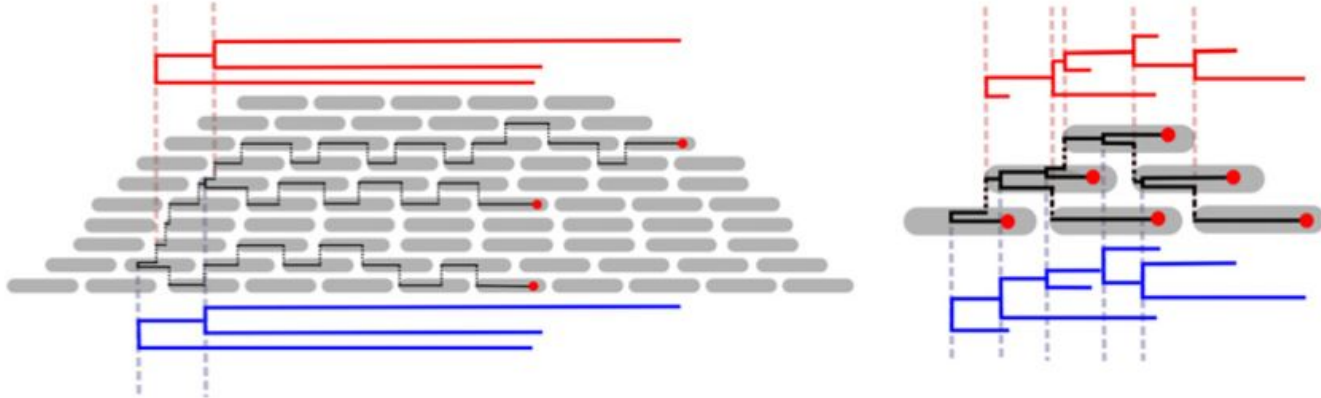
Within-host diversity

Incomplete transmission bottlenecks can lead to even more extreme discrepancies between transmission trees and phylogenies

De Maio *et al.* (PLoS Comp Bio, 2016)
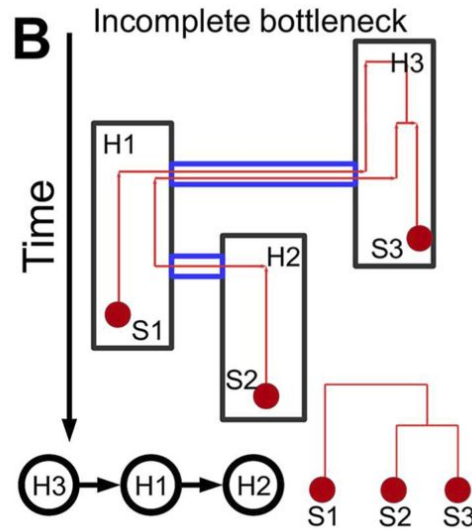
Within-host diversity

Multiple infections can cause hosts to be erroneously excluded from transmission chains.

De Maio *et al.* (PLoS Comp Bio, 2016)

Within-host diversity

But on the positive side, within-host diversity can also help link infections and resolve the directionality of transmission between a donor and recipient.

Direct transmission

Unsampled intermediate

Common source

Romero-Severson *et al.* (PNAS, 2016)

Phyletic relationships

The **phyletic relationships** among sampled pathogens can provide information about the source of transmission if we have multiple samples from each host.

TAXON 1 (monophyletic) TAXON 2 (polyphyletic) TAXON 3 (paraphyletic)

©1999 Addison Wesley Longman, Inc.

Within-host diversity

Let's consider the different phyletic relationships among lineages samples from the transmission pair A-B:

Direct transmission

Unsampled intermediate

Common source

Romero-Severson *et al.* (PNAS, 2016)
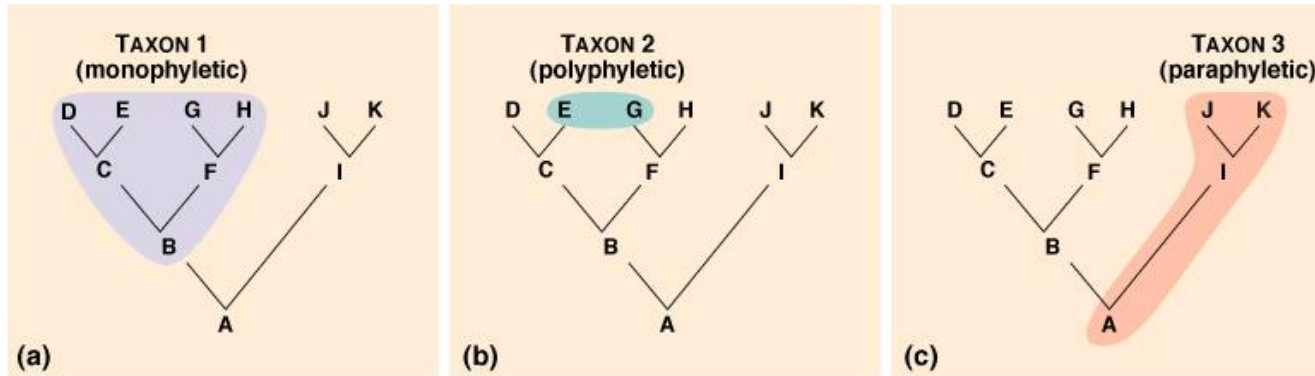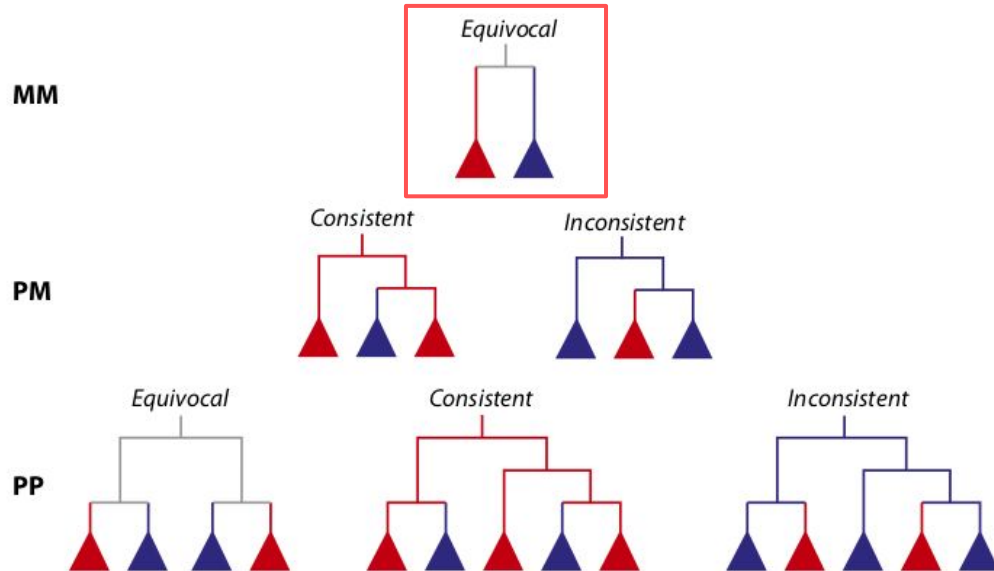
Phyletic relationships

The **phyletic relationships** among sampled lineages can provide information about the source of transmission if we have multiple samples from each host.

Monophyletic-Monophyletic (MM): Equivocal about the directionality of transmission, but likely to result from a common source of transmission

Romero-Severson *et al.* (PNAS, 2016)
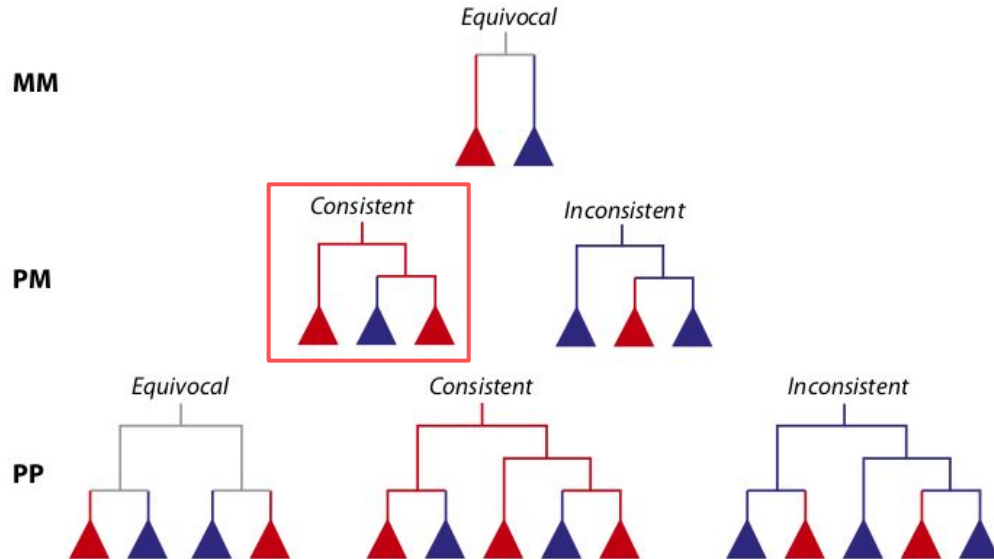
Phyletic relationships

The **phyletic relationships** among sampled lineages can provide information about the source of transmission if we have multiple samples from each host.

Paraphyletic-Monophyletic (PM): Donor is generally paraphyletic (red) while the recipient (blue) is monophyletic. Most likely results from direct or indirect transmission.

Romero-Severson *et al.* (PNAS, 2016)
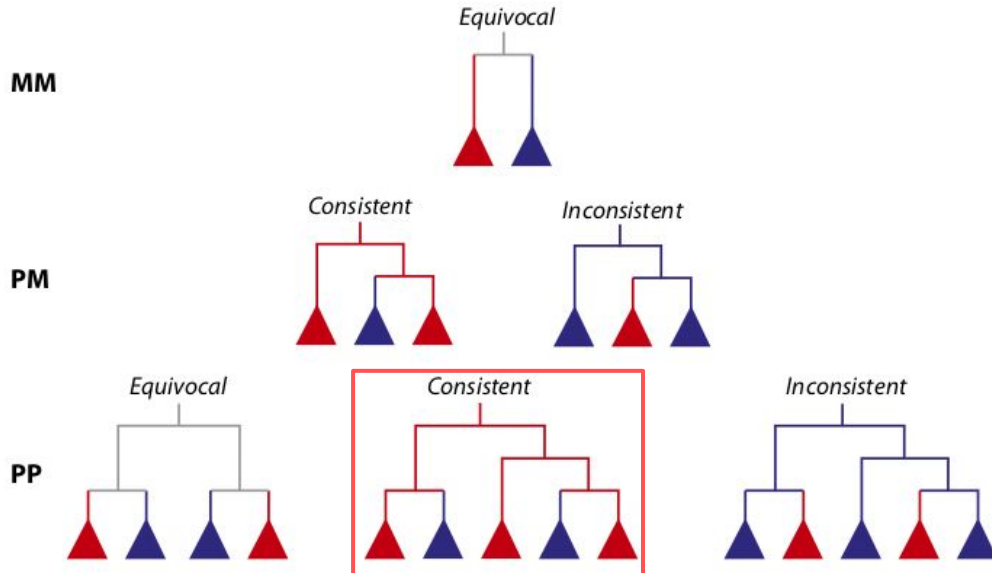
Phyletic relationships

The **phyletic relationships** among sampled lineages can provide information about the source of transmission if we have multiple samples from each host.

Paraphyletic-Polyphyletic (PP): Generally indicates direct transmission between donor (paraphyletic) and recipient (polyphyletic). Indirect transmission very improbable.

Romero-Severson *et al.* (PNAS, 2016)

Two main approaches

1. Methods that directly estimate the underlying transmission tree

2. Methods that reconstruct pathogen phylogenies and then infer transmission events between hosts
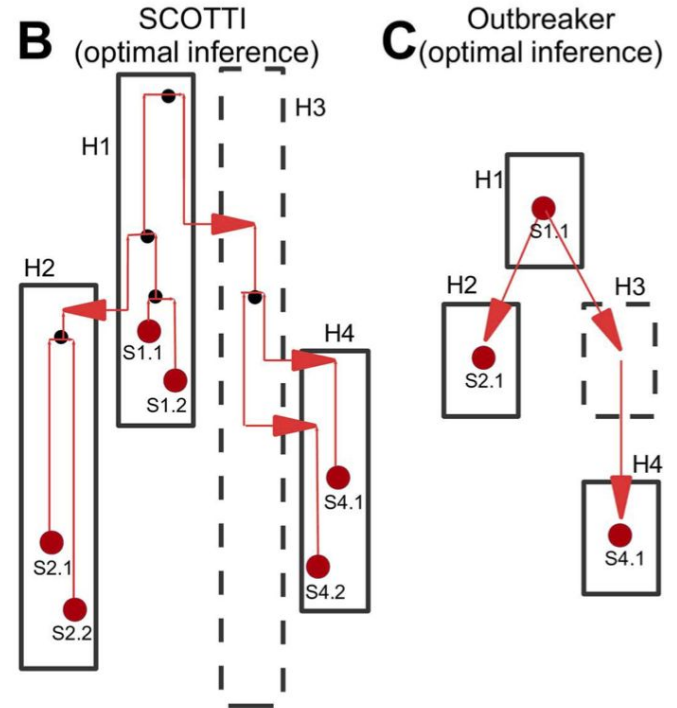
The SCOTTI Approach
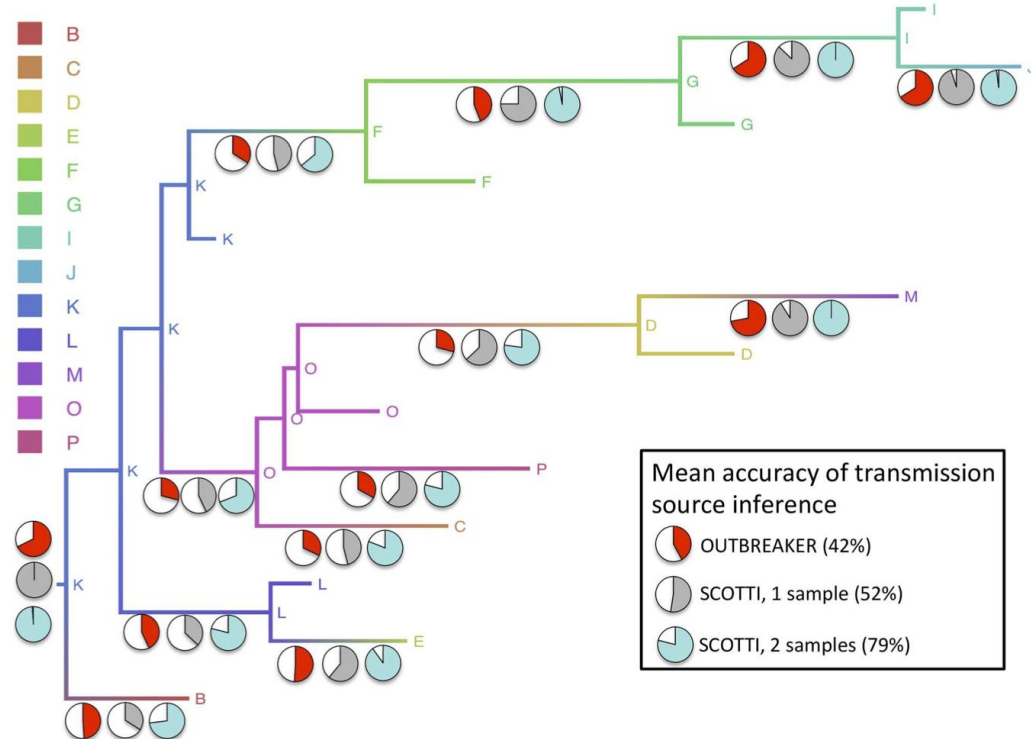
Structured COalescent Transmission Tree Inference

Treats each host as a different subpopulation in a structured coalescent model.

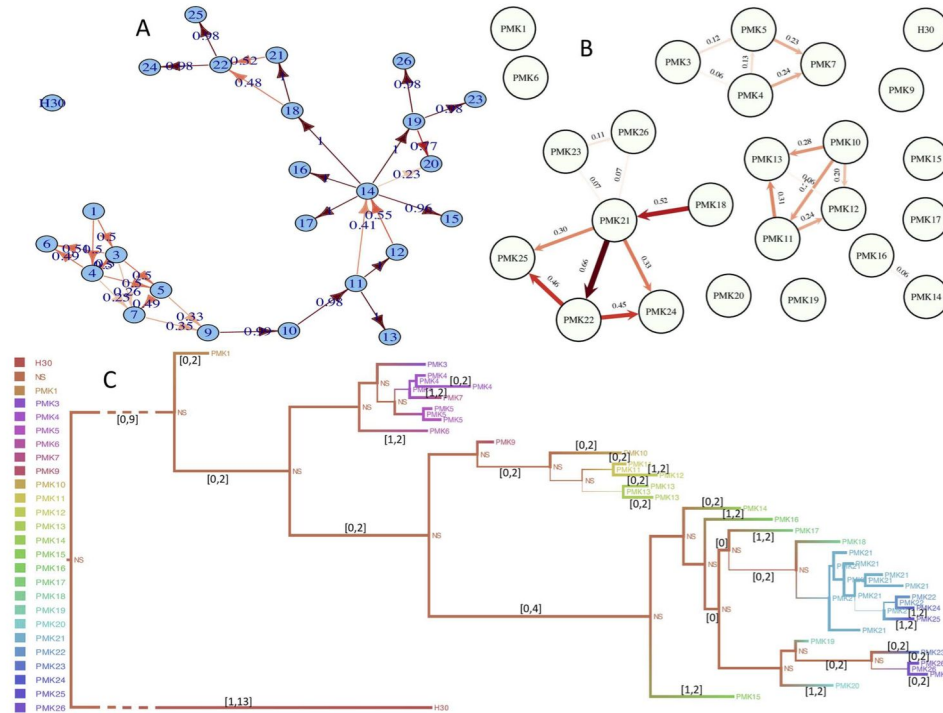Inferred migration events can be used to reconstruct transmission routes

Accounts for within-host diversity, unsampled hosts and incomplete transmission bottlenecks
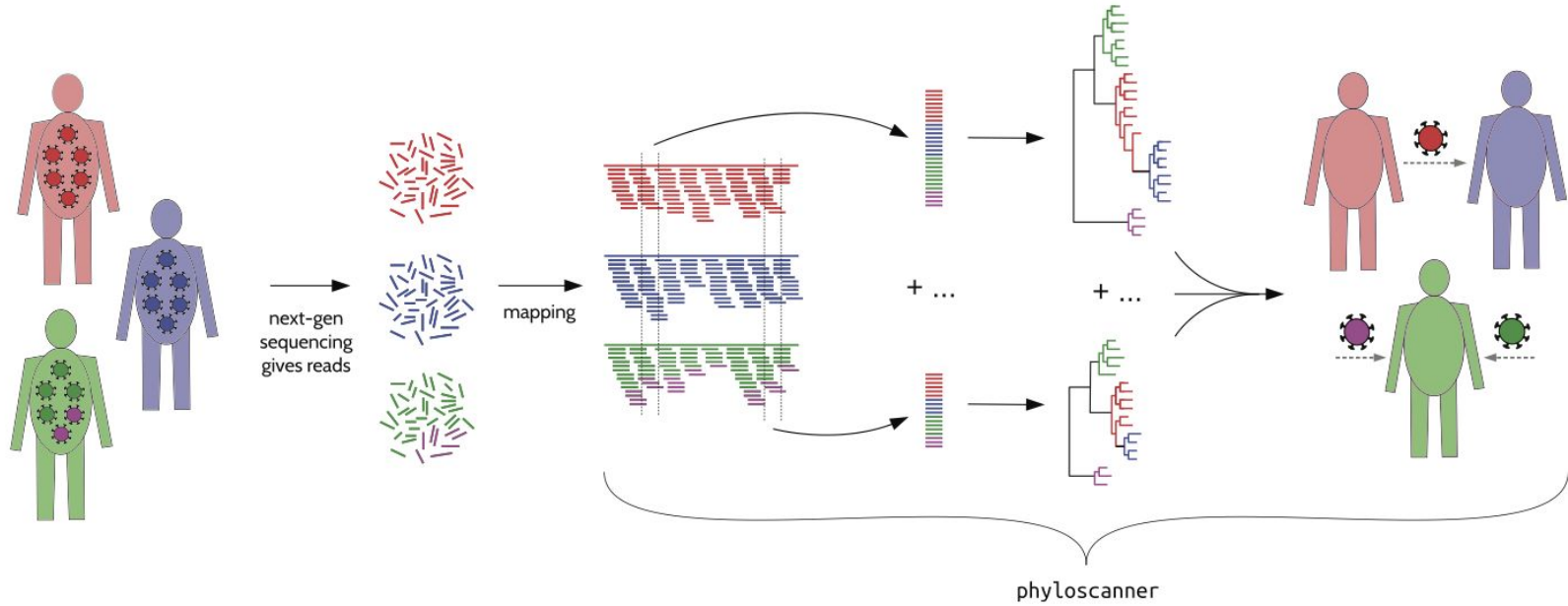

De Maio *et al.* (PLoS Comp Bio, 2016)

SCOTTI versus Outbreaker

Mean accuracy of transmission source inference
- OUTBREAKER (42%)
- SCOTTI, 1 sample (52%)
- SCOTTI, 2 samples (79%)

De Maio *et al.* (PLoS Comp Bio, 2016)

Klebsiella outbreak reconstruction

De Maio *et al.* (PLoS Comp Bio, 2016)

The phyloscanner approach
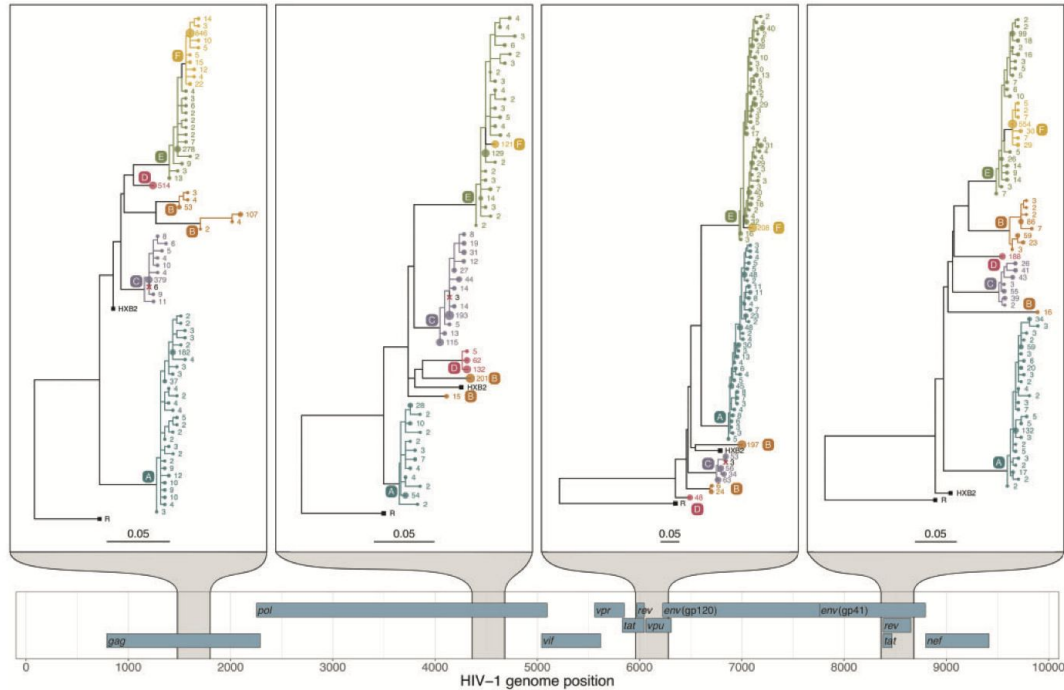

Wymant *et al.* (MBE, 2017)

The phyloscanner approach

Wymant *et al.* (MBE, 2017)

Summary

We can reconstruct transmission trees directly from genetic data or in combination with additional epidemiological data

Reconstructing transmission trees from genetic data alone is very difficult especially if there are many unsampled hosts and high within-host genetic diversity

Newer (phylogenetic) approaches leverage the ability to sequence multiple pathogens from each host to more accurately reconstruct transmission chains.