

# **Inferring transmission trees and who's infecting whom**

Molecular Epidemiology of Infectious Diseases  
Lecture 5

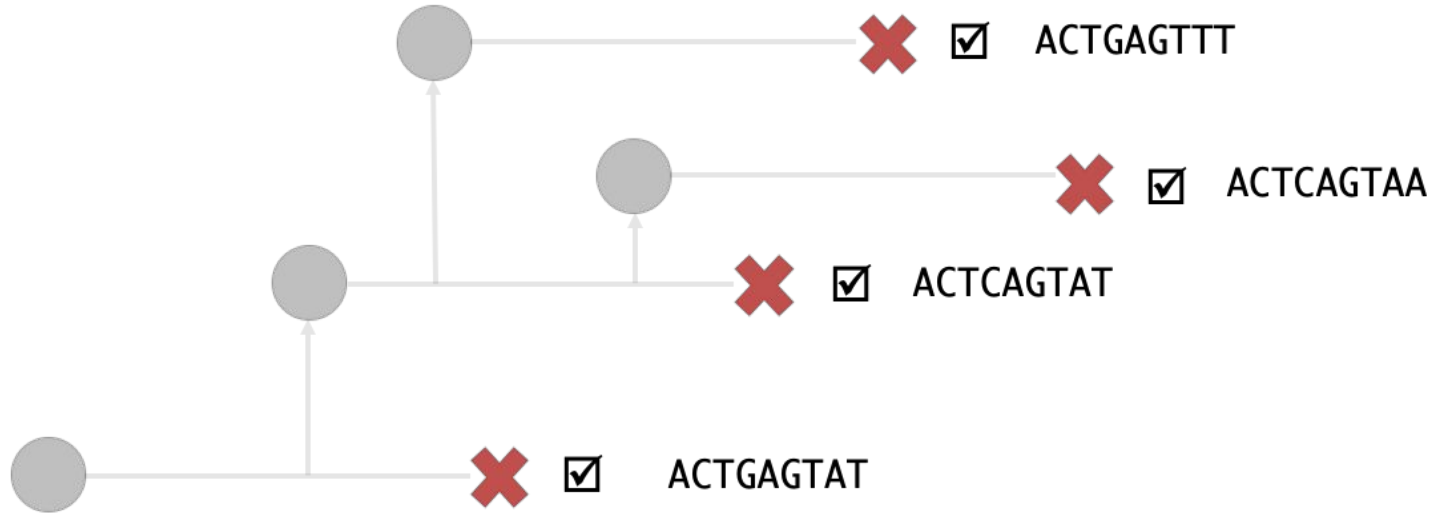
February 14th, 2022

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

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

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# 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

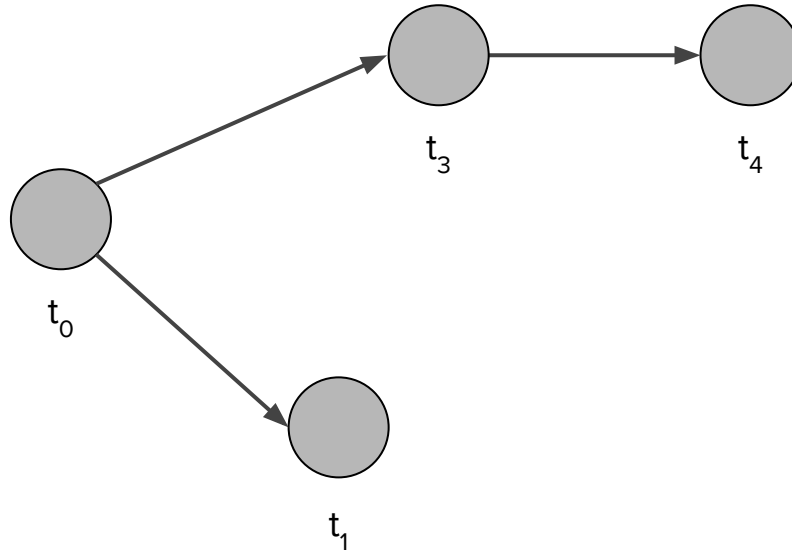
1. Methods that directly estimate the underlying transmission tree
2. Methods that reconstruct pathogen phylogenies and then infer transmission routes between hosts

# 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

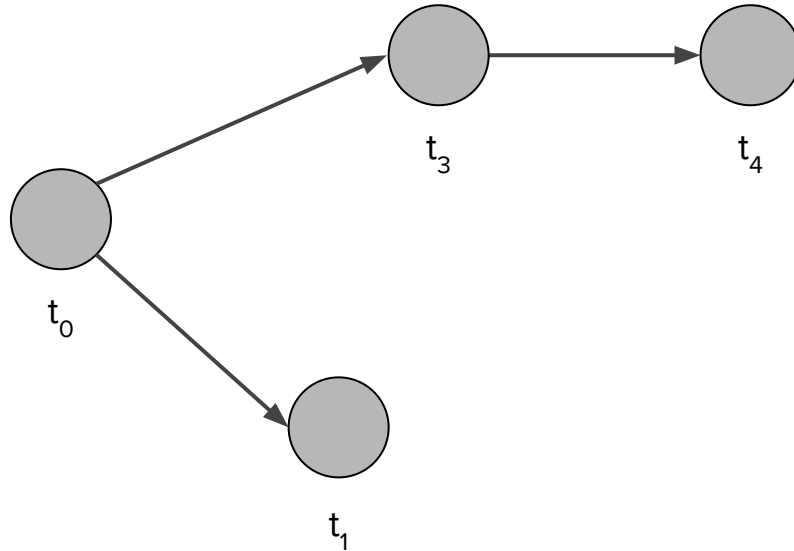
General goal is to probabilistically reconstruct likely transmission links





# Transmission tree reconstruction

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



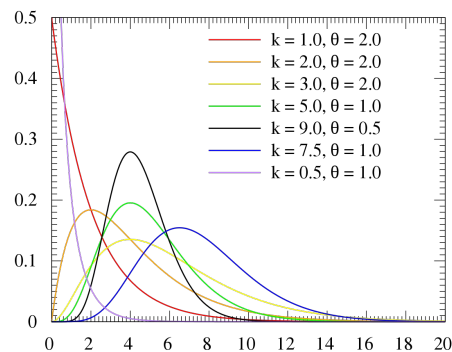
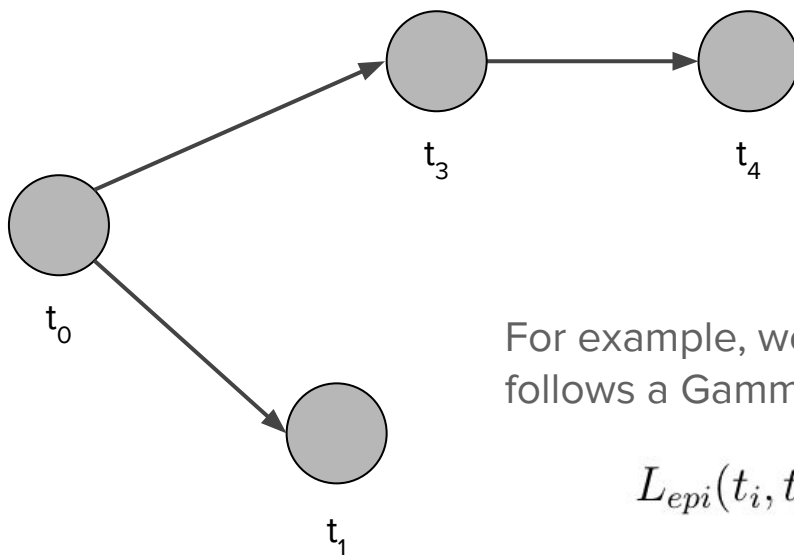
# Transmission tree reconstruction

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

# Example: 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:

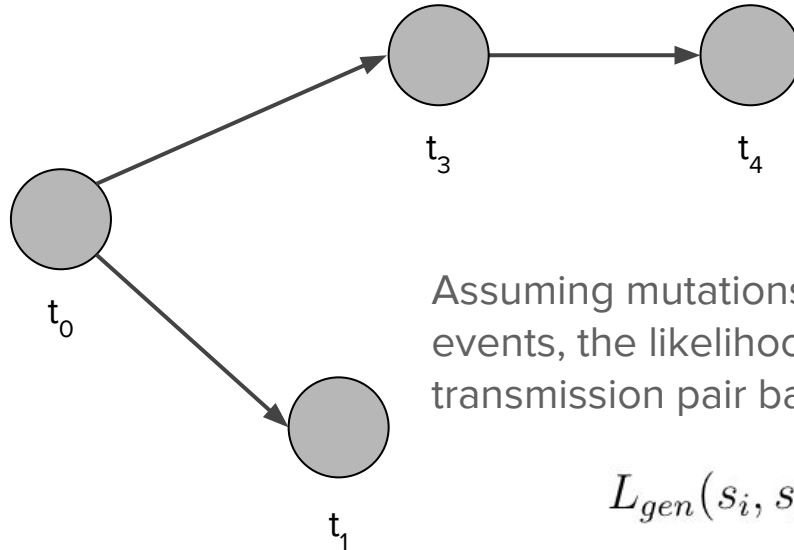


For example, we could assume the generation time follows a Gamma distribution:

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

# 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:

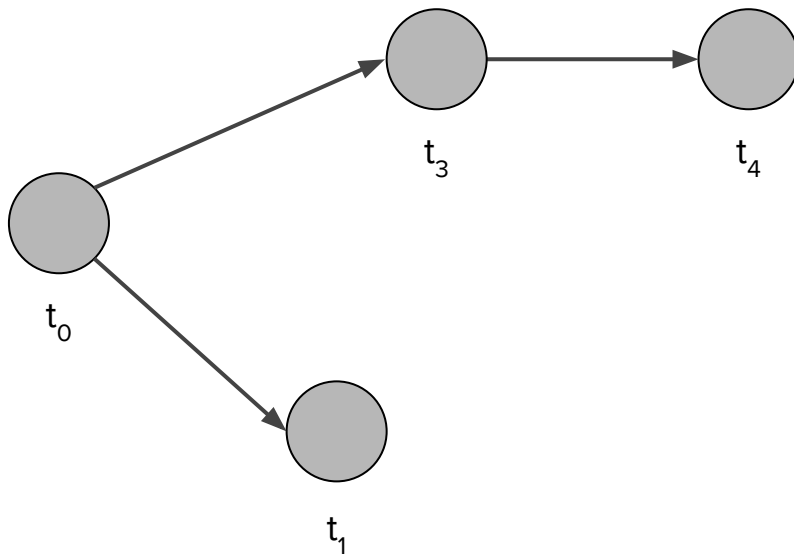


Assuming mutations happen at rate  $\mu$  at transmission events, the likelihood of two sequences being in a transmission pair based on their genetic distance  $d(s_i, s_j)$ :

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

# Transmission tree reconstruction

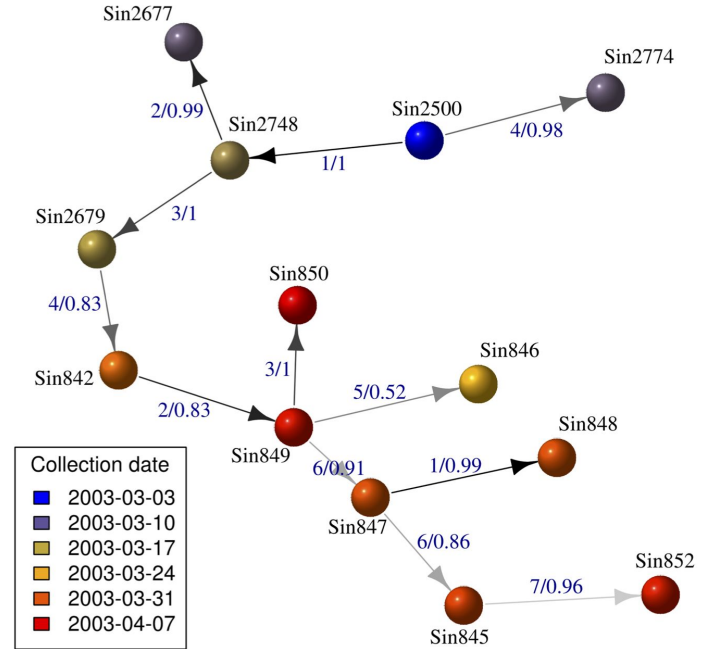
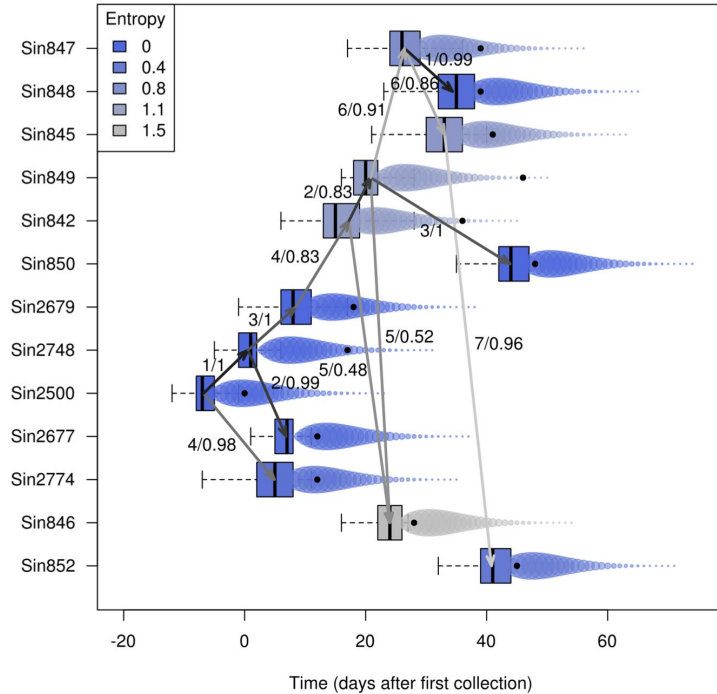
Our goal is to find the transmission tree that maximizes the **overall likelihood** of the infection times and sequence data across all transmission pairs:



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



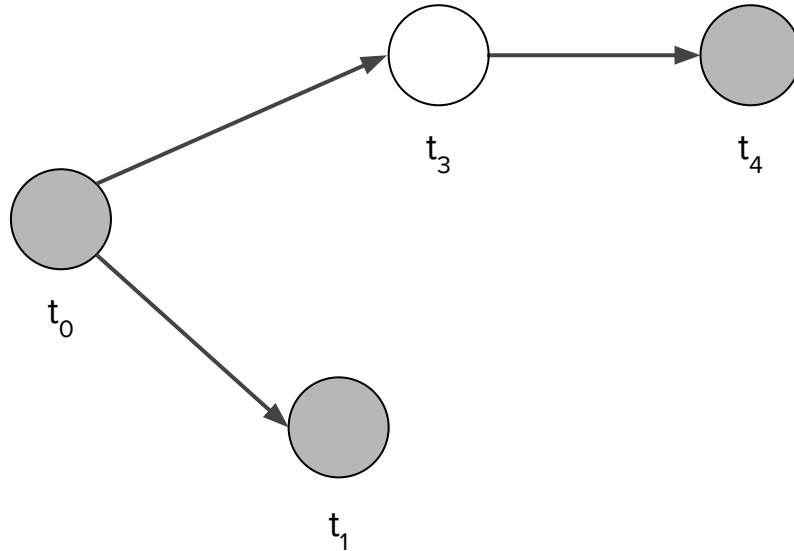
# Direct transmission tree reconstruction

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

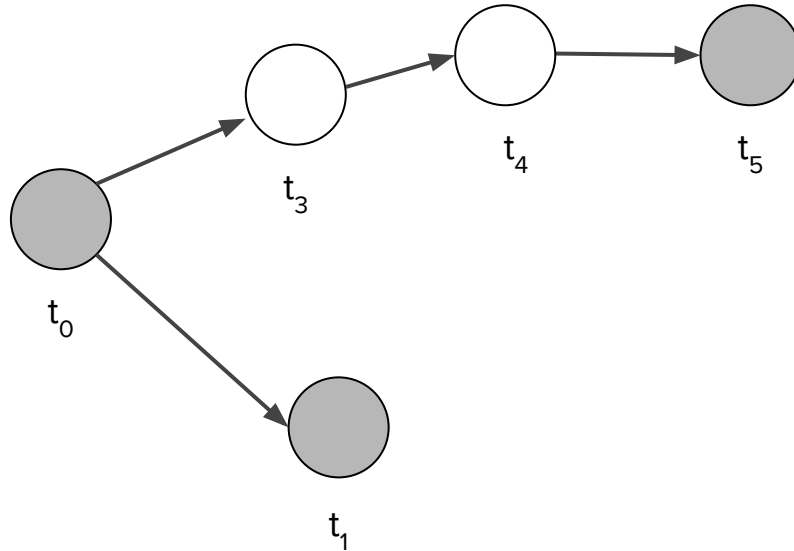
The problem is that we generally have incomplete sampling with at least some unobserved infections.





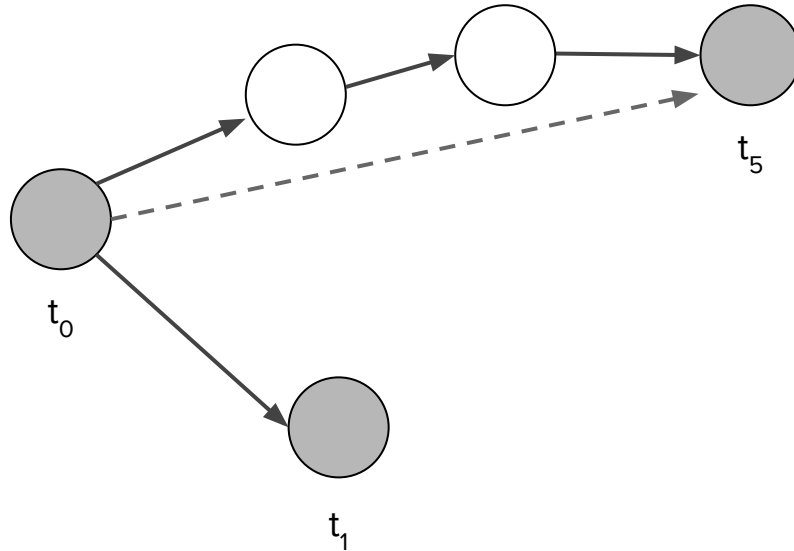
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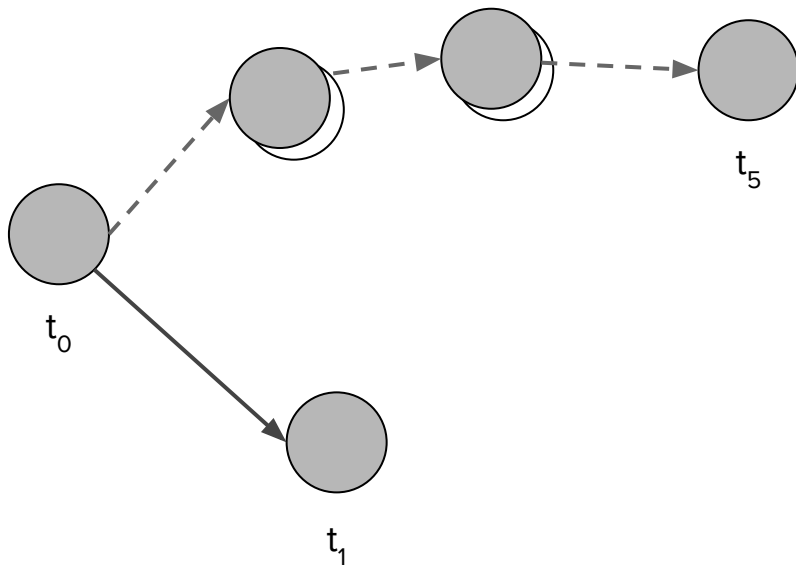
# Transmission tree reconstruction

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



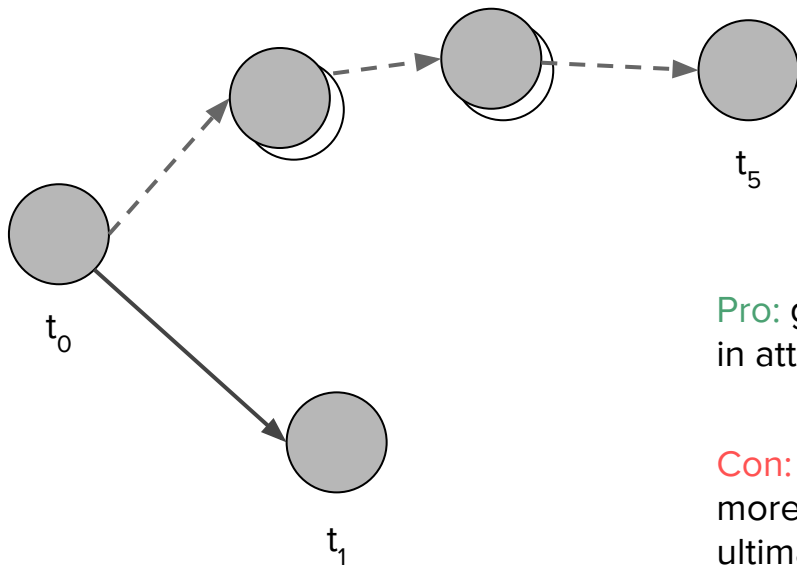
# Data augmentation

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



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**Pro:** guards against overconfidence in attributing sources.

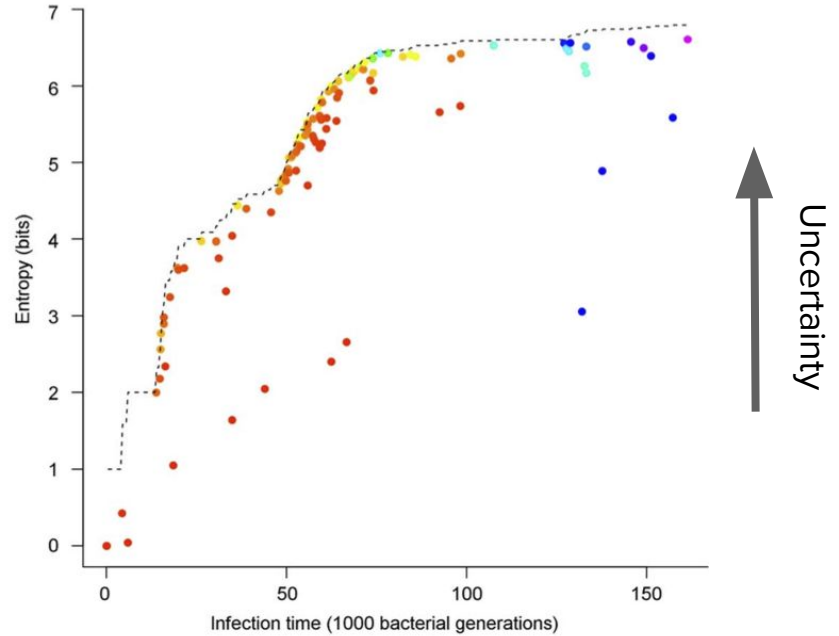
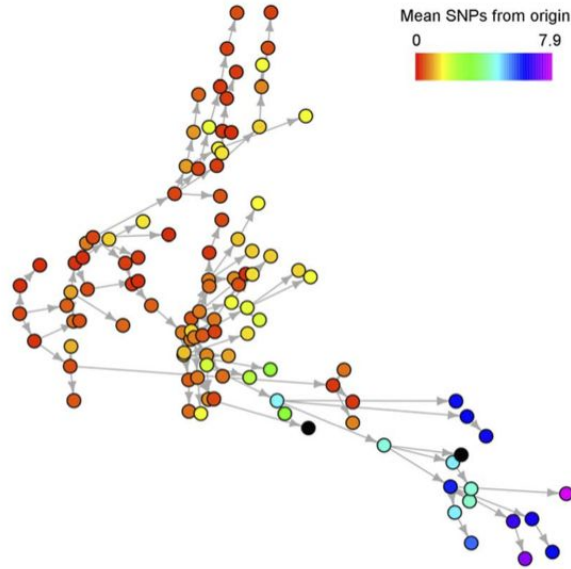
**Con:** uncertainty will only grow with more unsampled hosts so ultimately doomed to fail.

# Direct transmission tree reconstruction

Direct reconstruction generally works well when:

- Outbreaks are small and we can sample nearly all infected hosts
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# Effect of overlapping infections



# 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

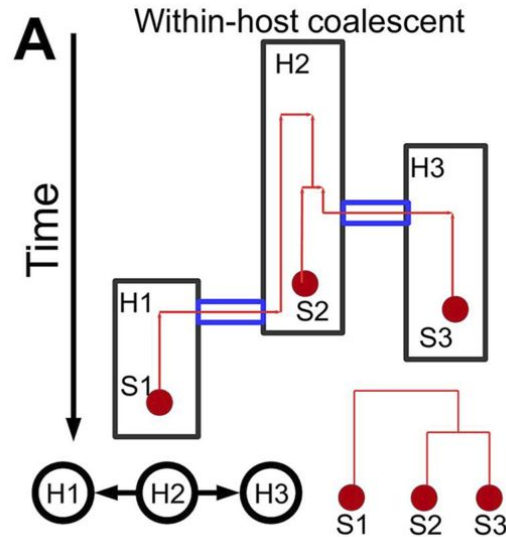
**So far we have  
completely ignored  
within-host genetic  
diversity!**





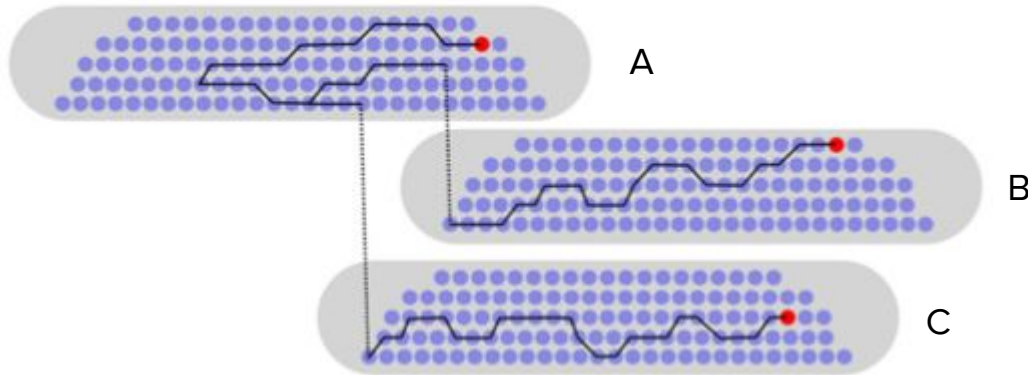
# Within-host diversity

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



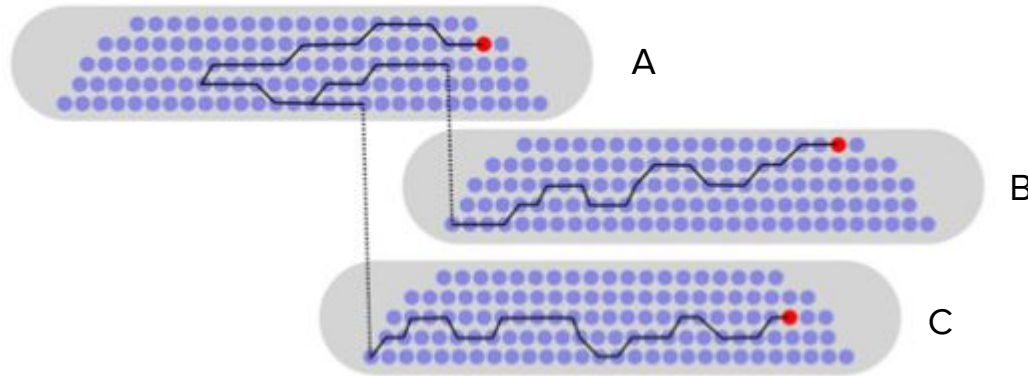
# Within-host diversity

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

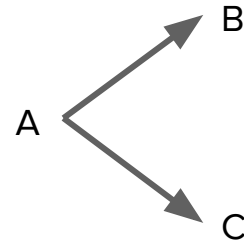


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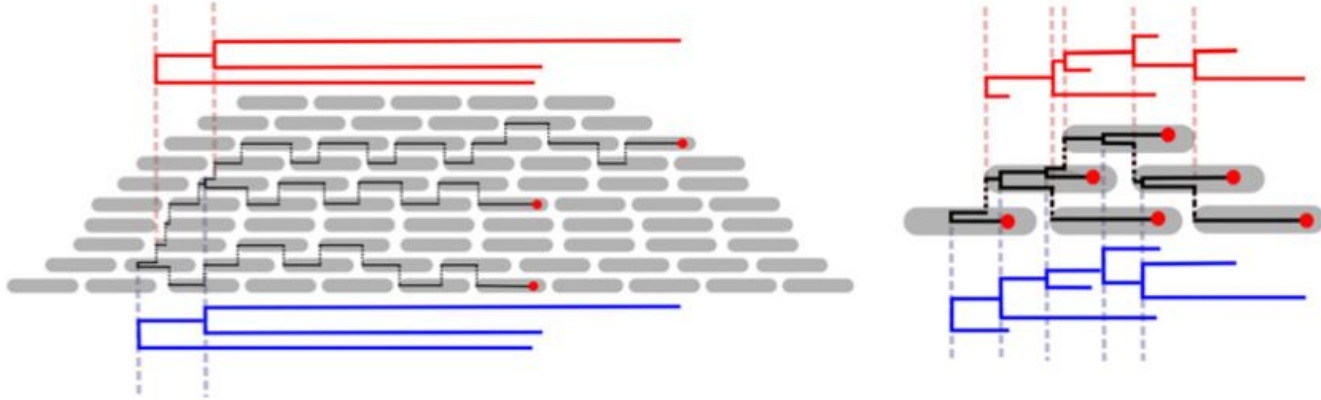
Actual transmission tree:



**But all three phylogenetic trees are possible!**

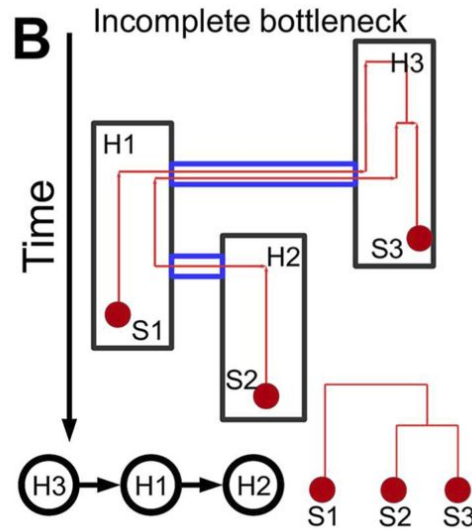
# Within-host diversity

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



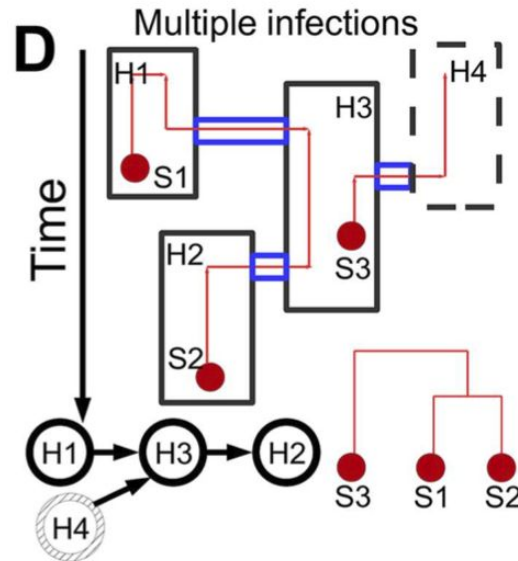
# Within-host diversity

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



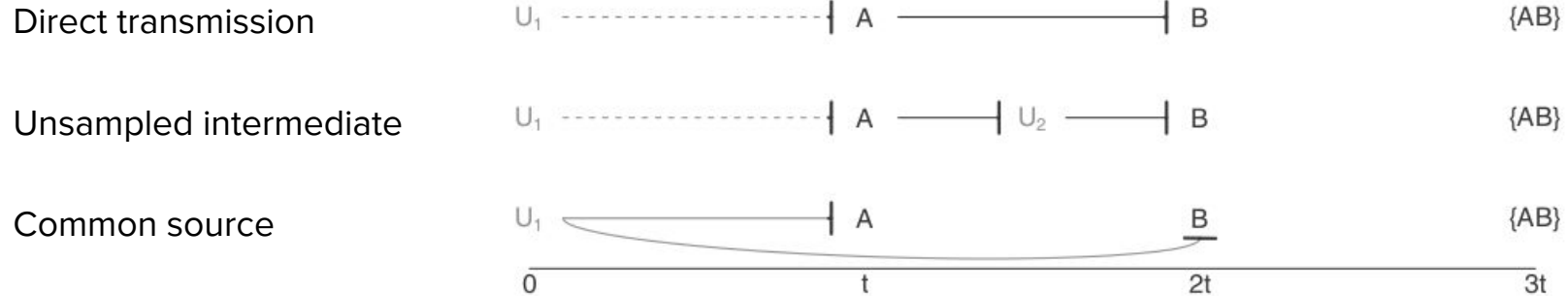
# Within-host diversity

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



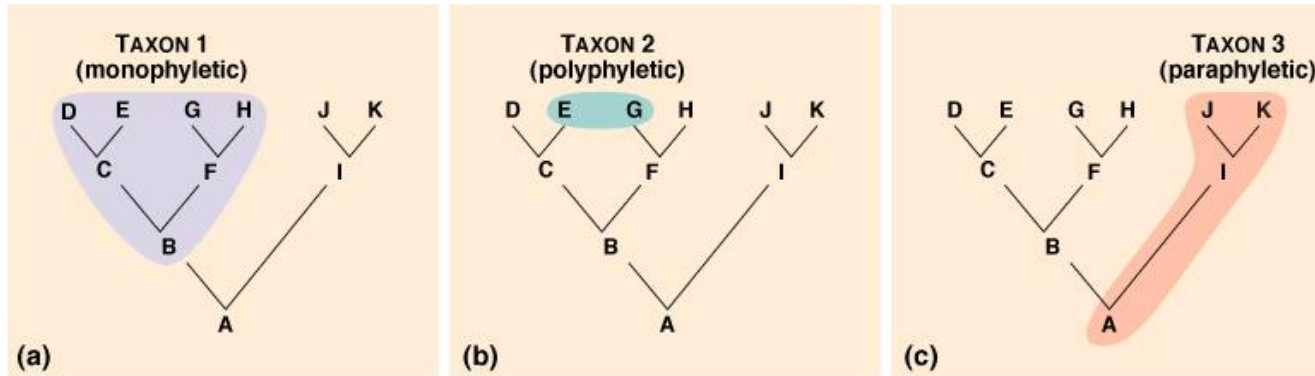
# Within-host diversity

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



# Phyletic relationships

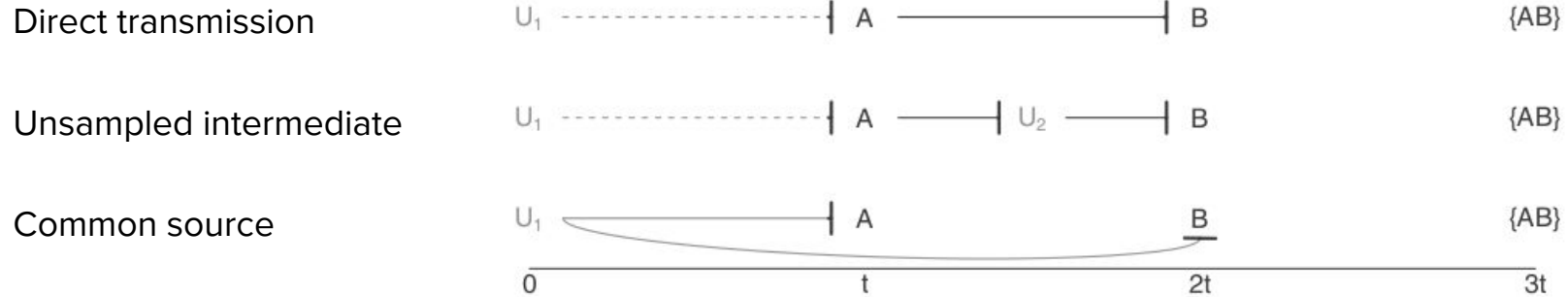
The **phyletic relationships** among sampled pathogens can provide information about the source of transmission if we have multiple samples from each host.





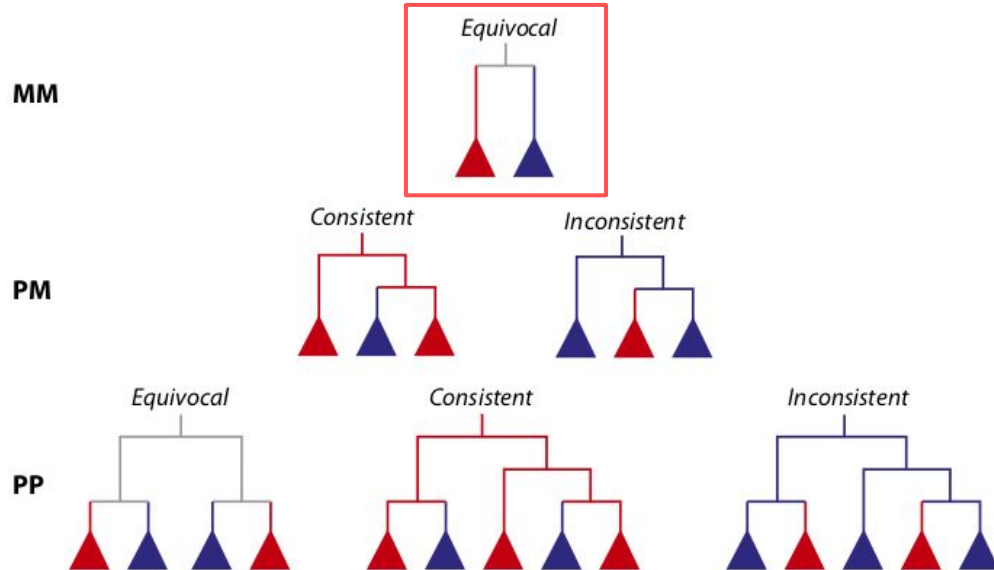
# Within-host diversity

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



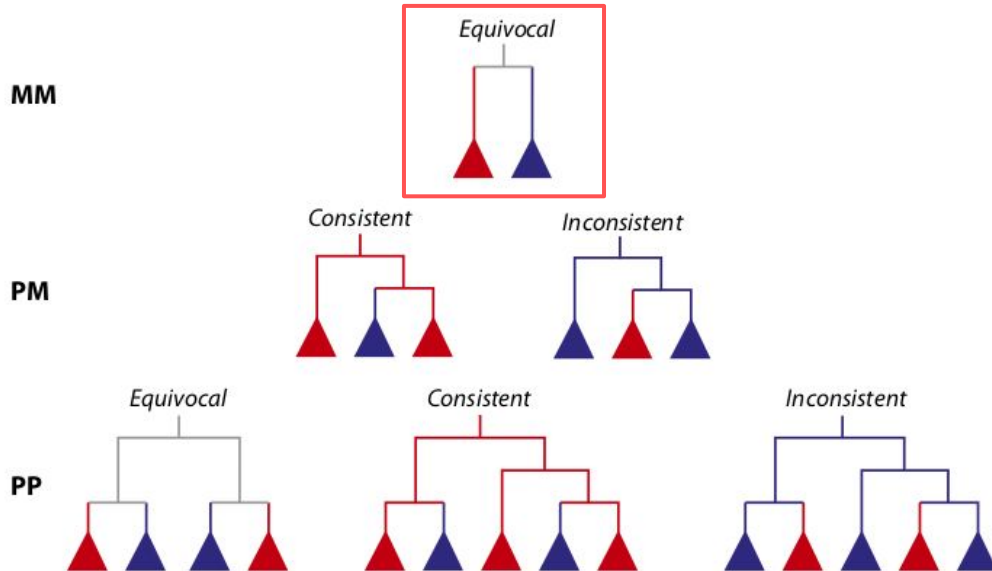
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# Phyletic relationships

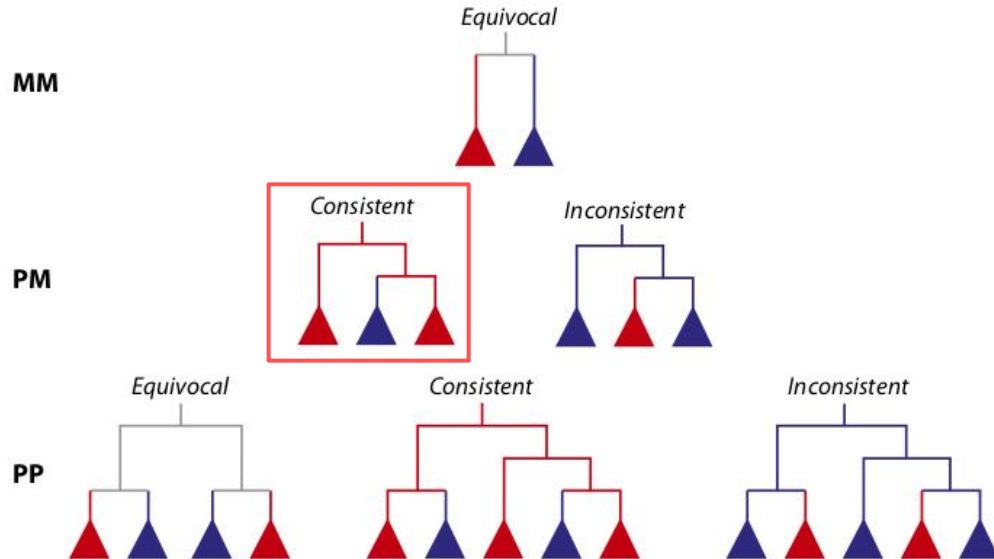
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**Monophyletic-Monophyletic (MM):** Equivocal about the directionality of transmission, but likely to result from a common source of transmission

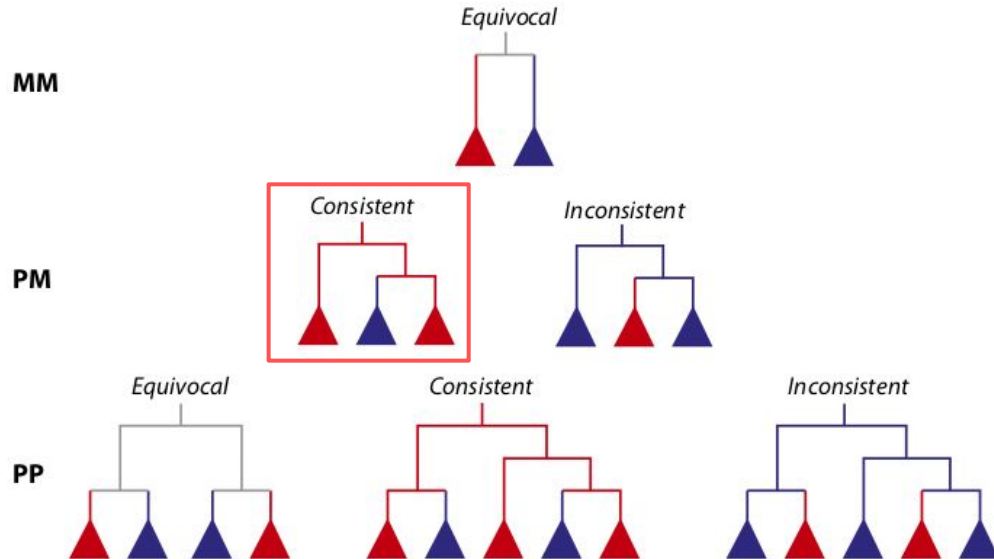
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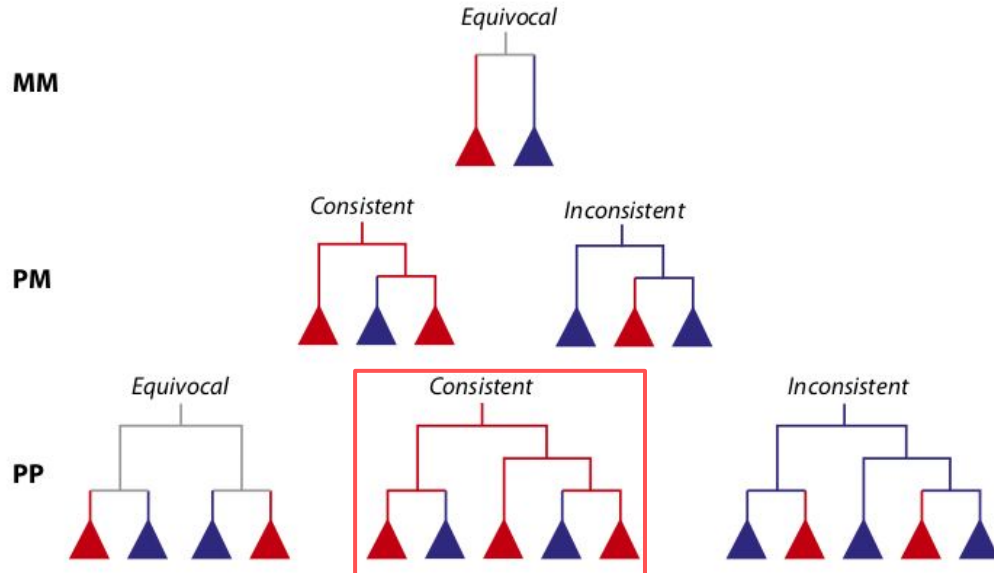
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**Paraphyletic-Monophyletic (PM):** Donor is generally paraphyletic (red) while the recipient (blue) is monophyletic. Most likely results from direct or indirect transmission.

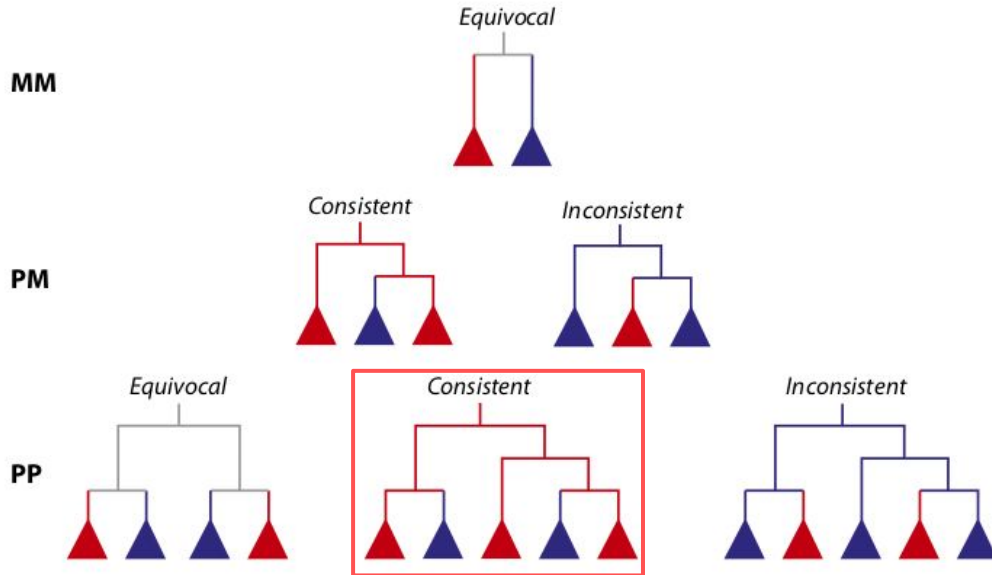
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## **Paraphyletic-Polyphyletic (PP):**

Generally indicates direct transmission between donor (paraphyletic) and recipient (polyphyletic). Indirect transmission very improbable.

# 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



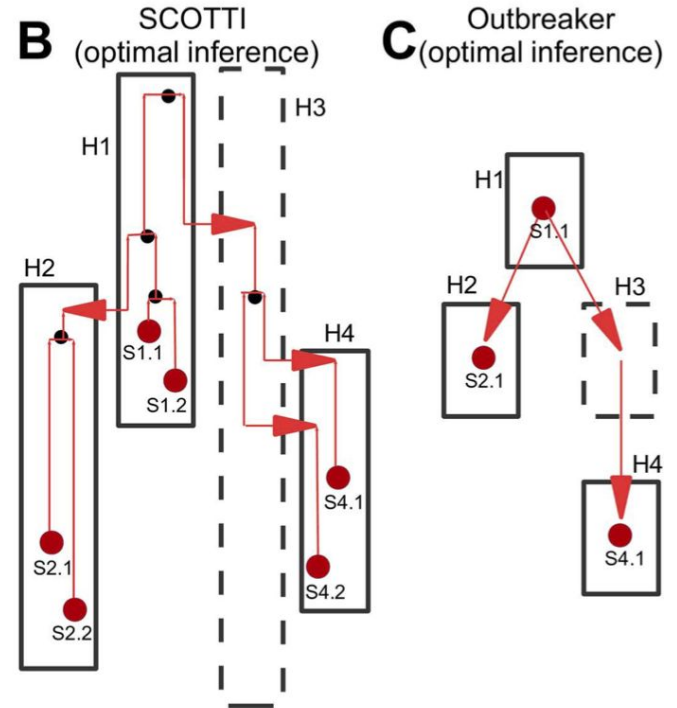
# The SCOTTI Approach

## Structured COalescent Transmission Tree Inference

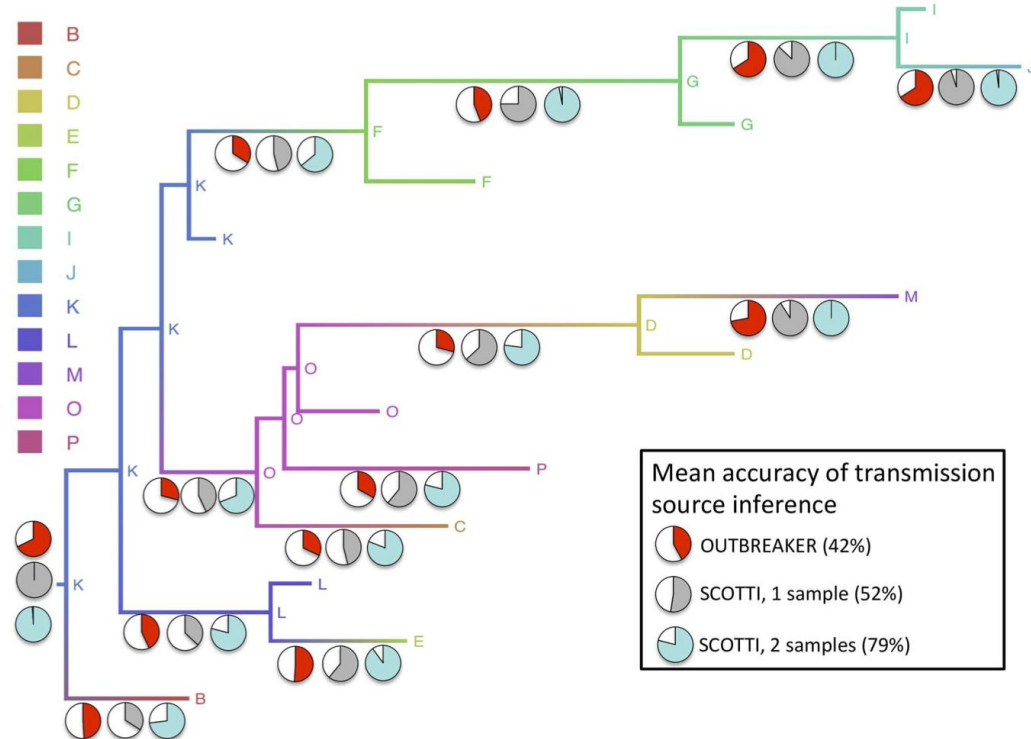
Treats each host as a different subpopulation in a structured coalescent model.

Inferred migration events can be used to reconstruct transmission routes

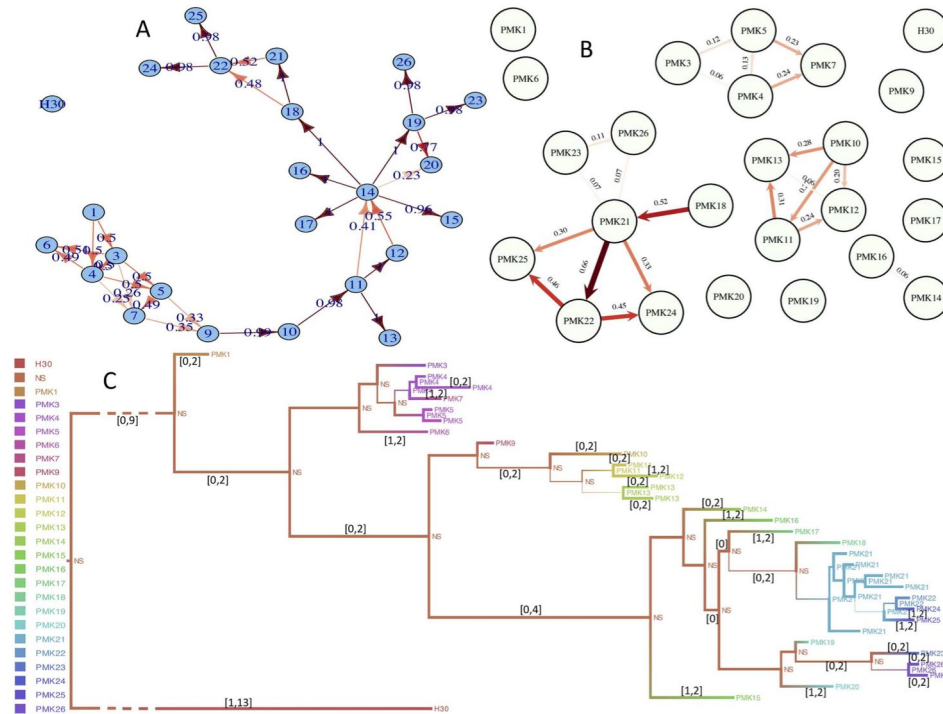
Accounts for within-host diversity, unsampled hosts and incomplete transmission bottlenecks



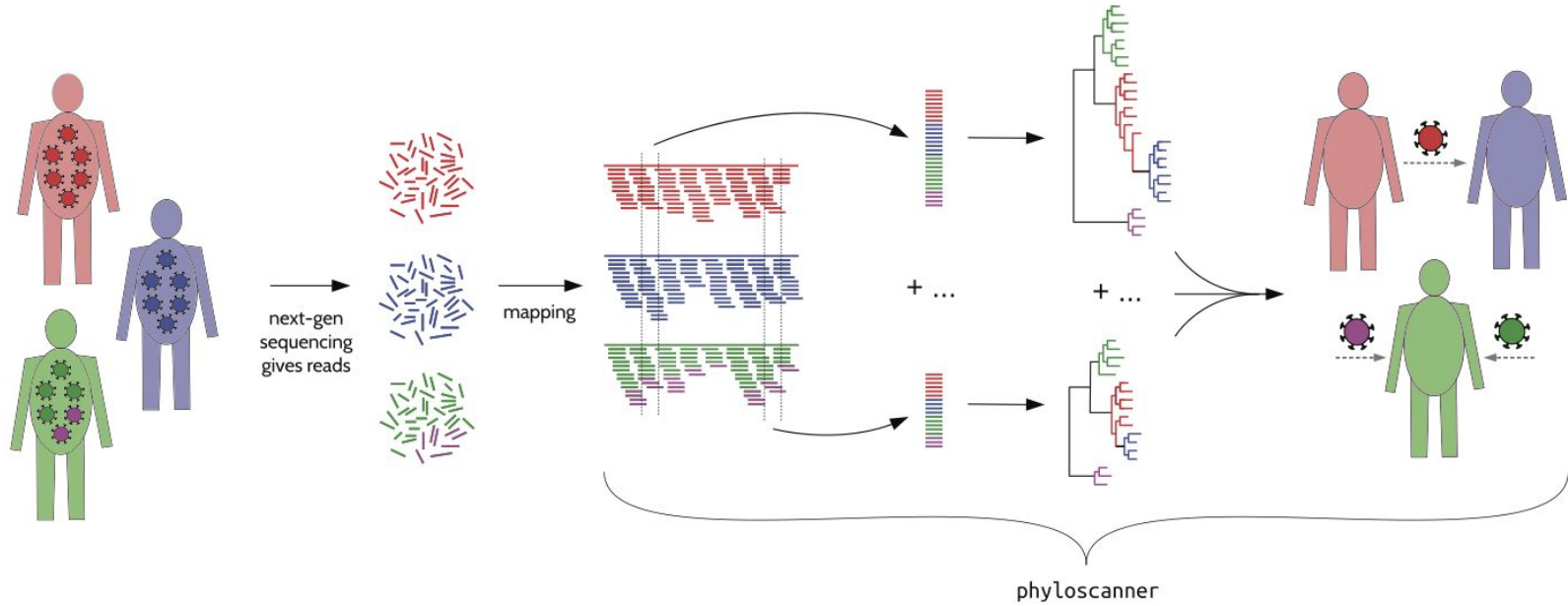
# SCOTTI versus Outbreaker



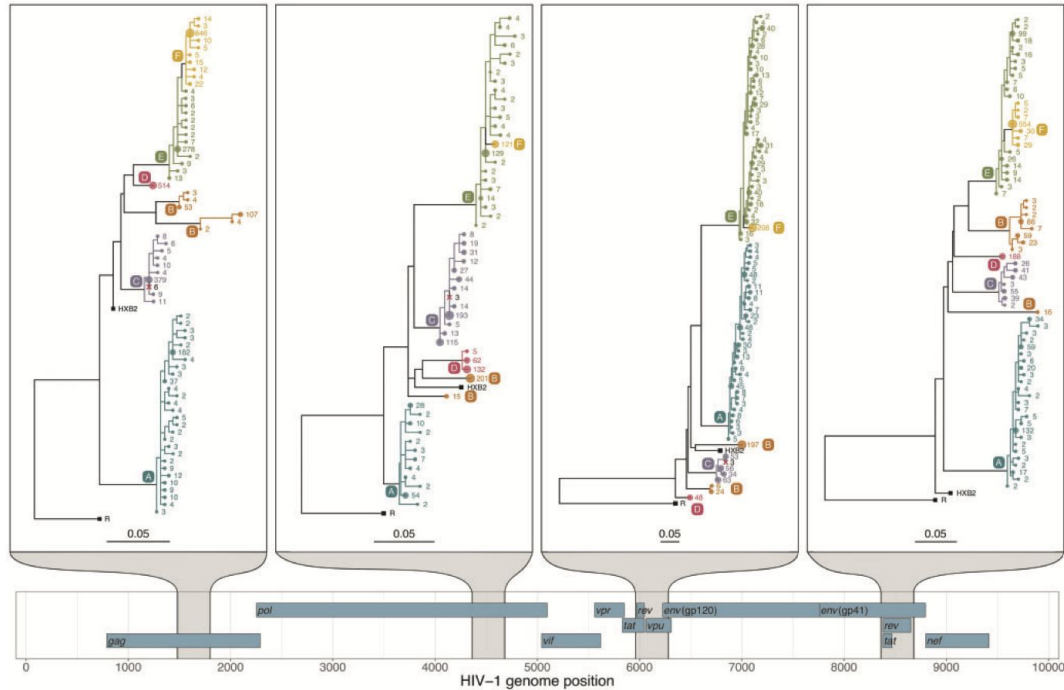
# *Klebsiella* outbreak reconstruction



# The phyloscanner approach



# 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.