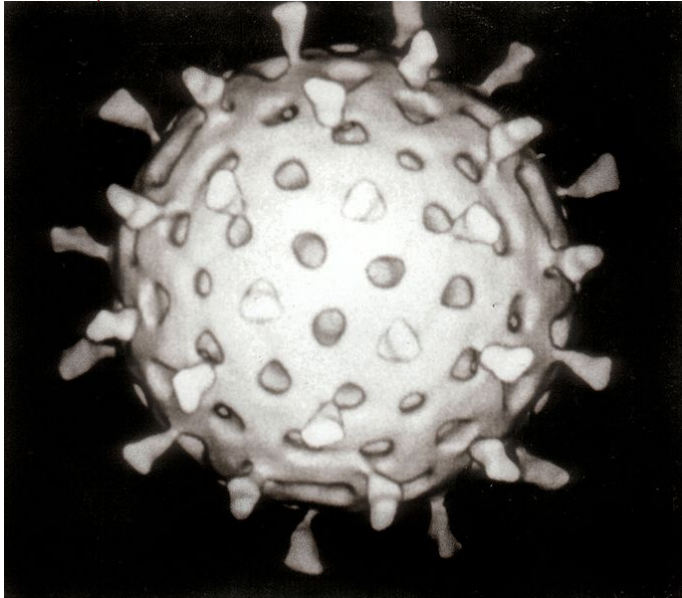
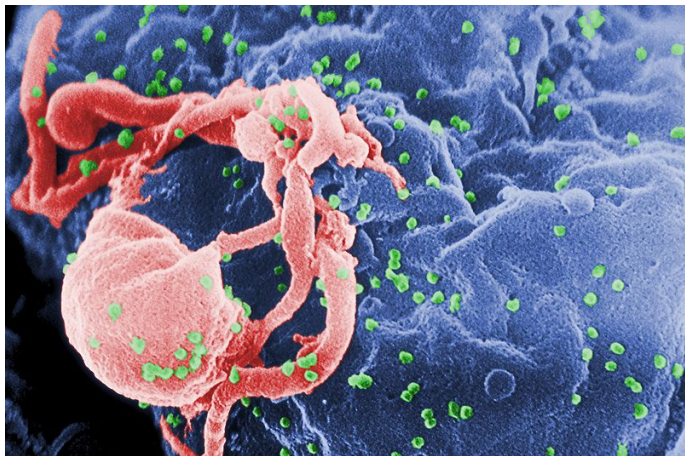


Bacteria, viruses and retroviruses evolve fast



Rate of evolution of HIV is maybe 6 orders of magnitude greater than in humans



See as much evolution in an HIV in a few years as in entire human history



Figure S3 (a) PB2 segment

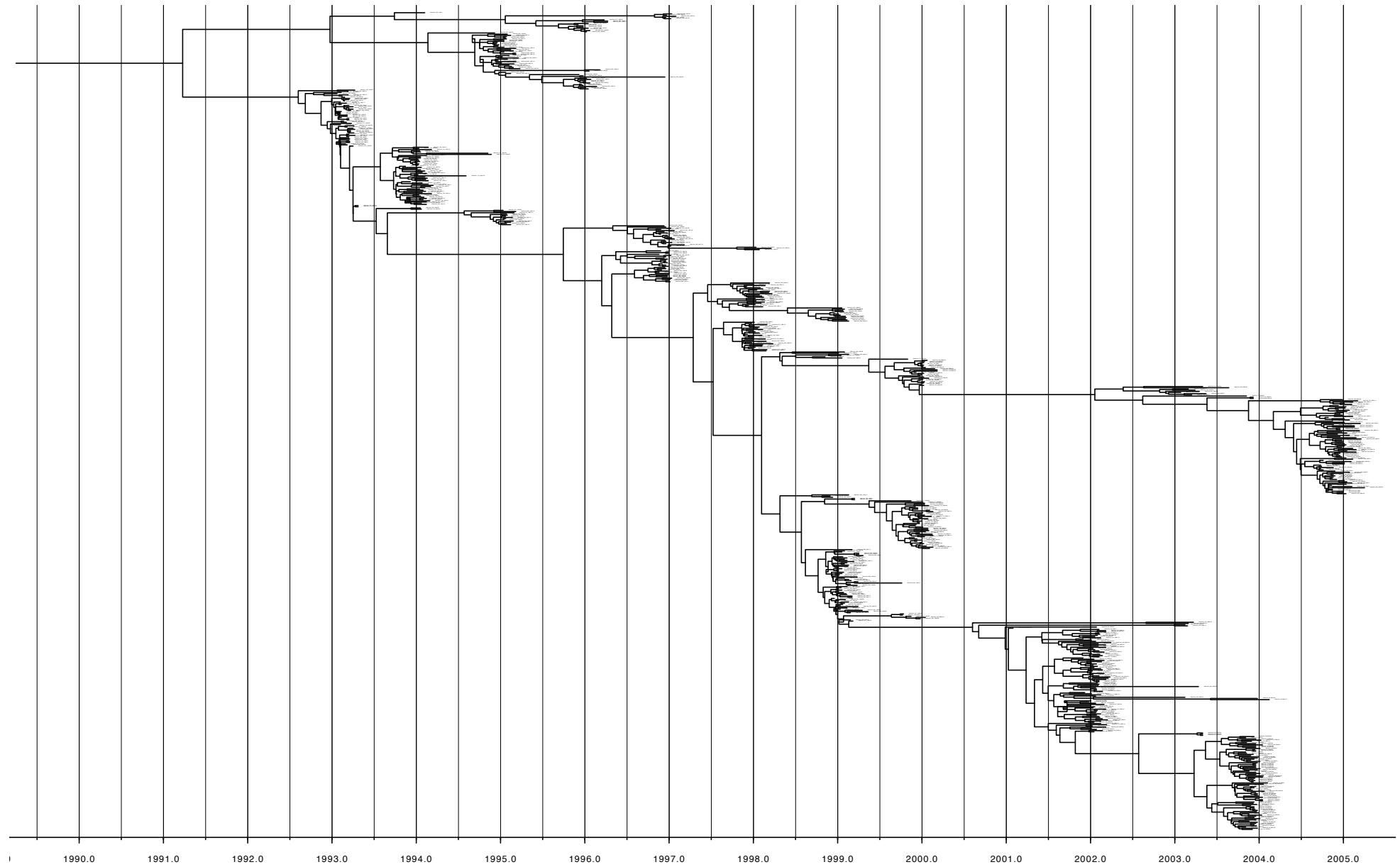
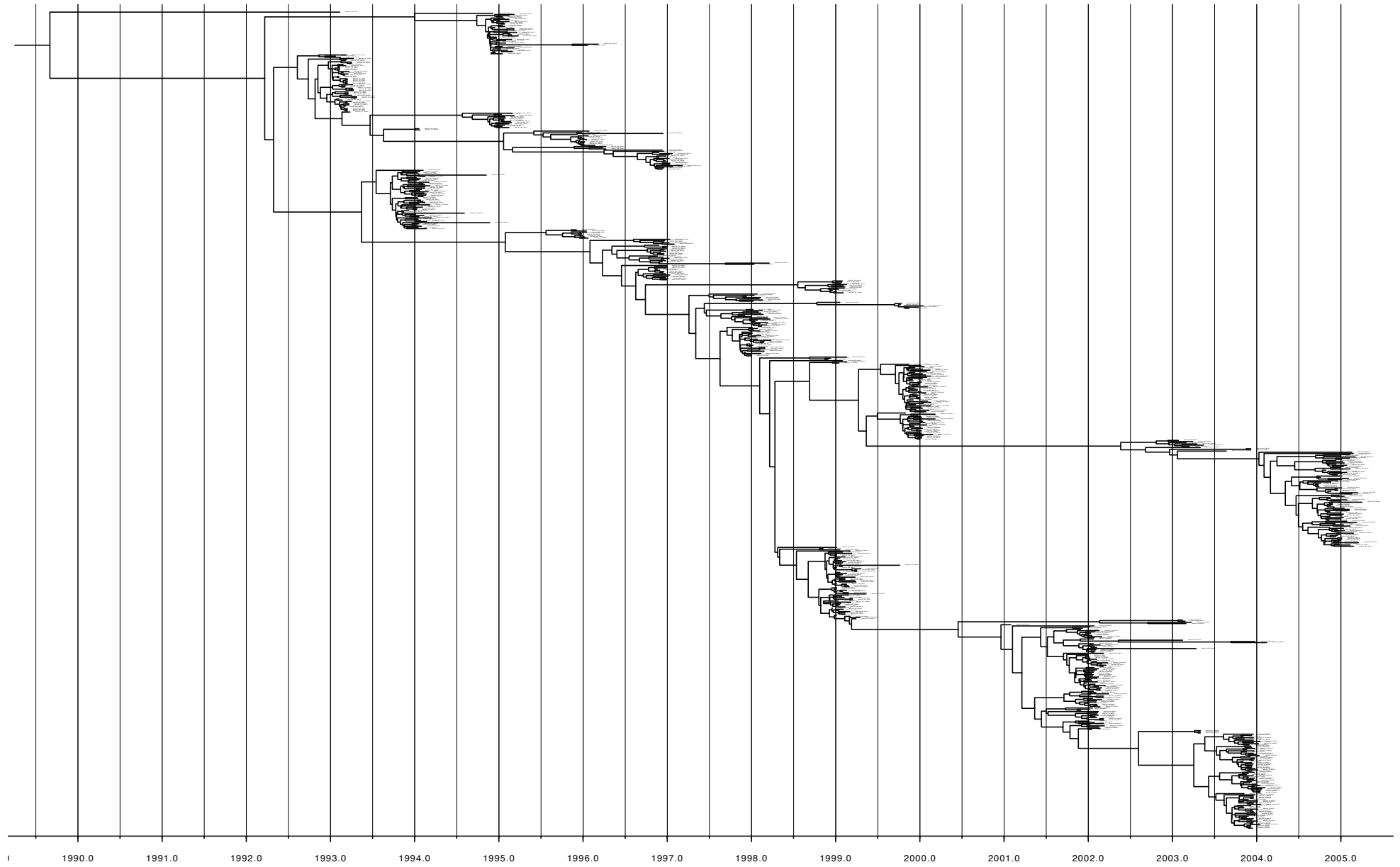
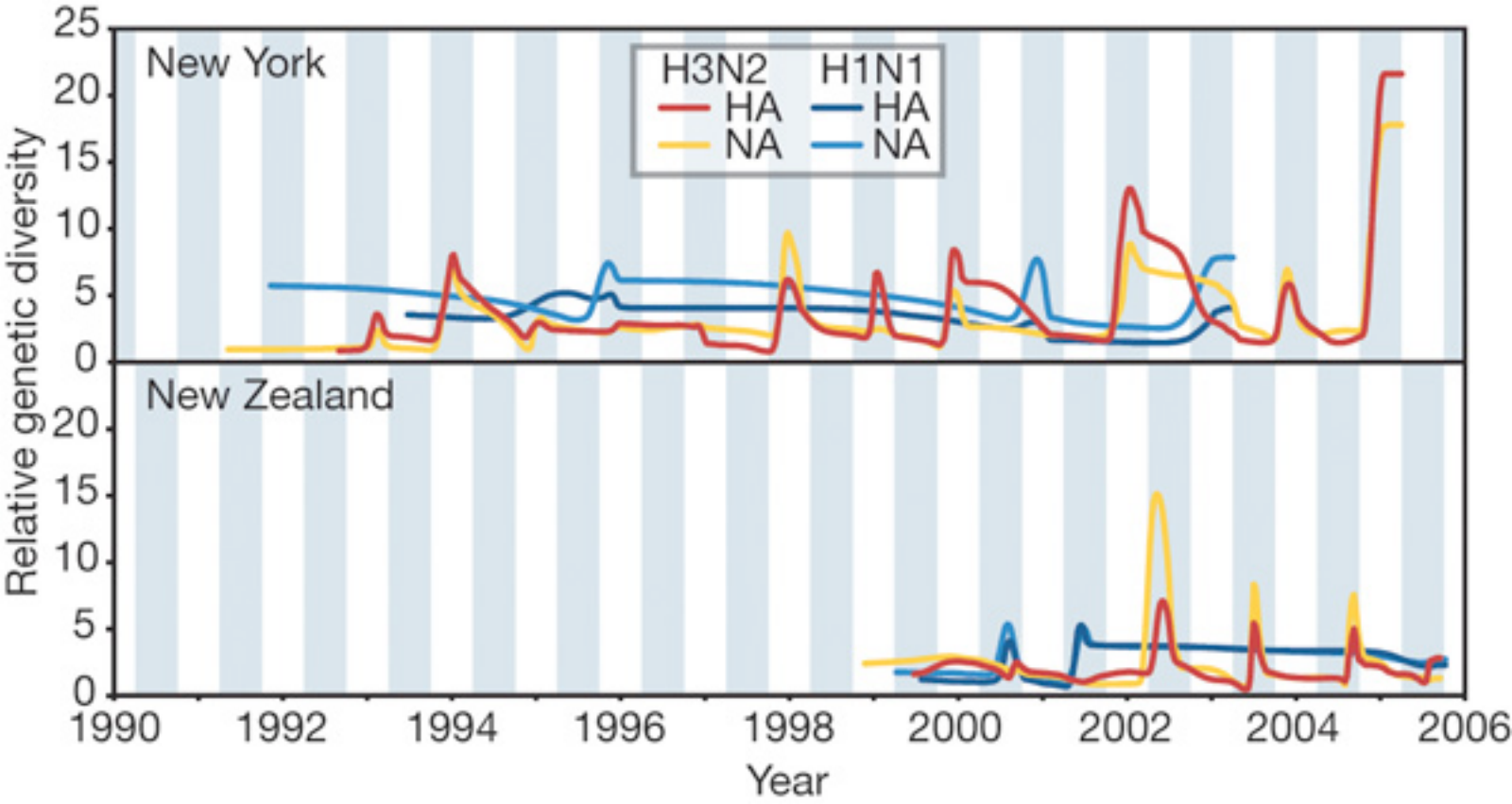
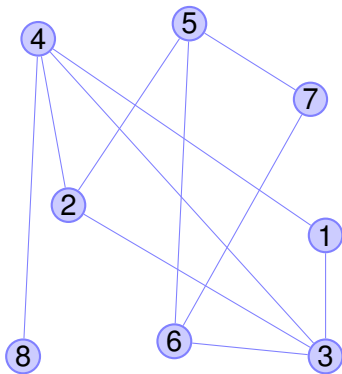


Figure S3 (b) PB1 segment



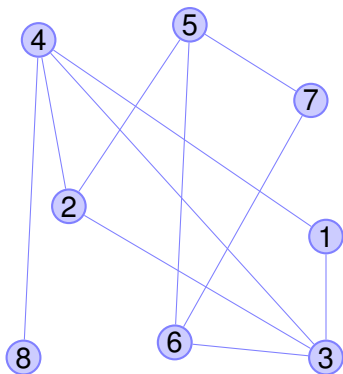


Networks are a natural way to model heterogeneity of contacts among hosts



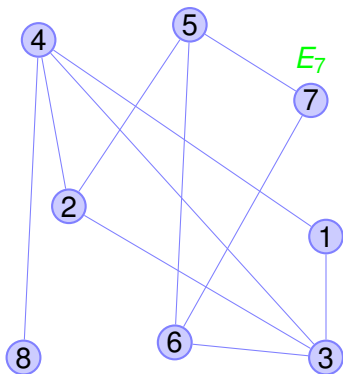
- ▶ Nodes represent hosts, edges represent contacts
- ▶ We want to use available data to estimate good models of networks

Contact networks and transmission trees



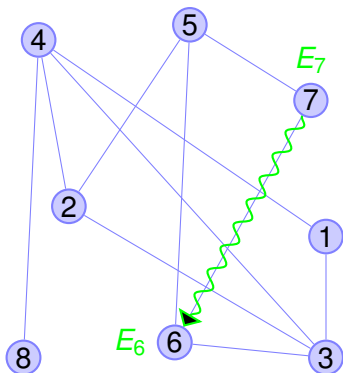
- Assume a *contact network* \mathcal{G} on N individuals: A “contact” is necessary for disease transmission.

Contact networks and transmission trees



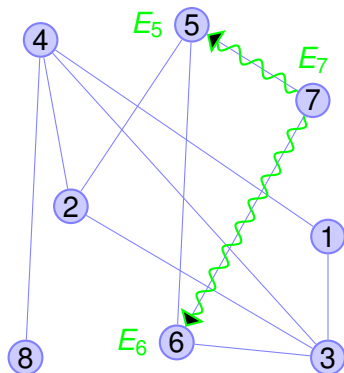
- ▶ Assume a *contact network* \mathcal{G} on N individuals: A “contact” is necessary for disease transmission.
- ▶ Beginning with the first infected, disease is spread at exponential rate β , defining a subtree of the contact network called the *transmission tree* \mathcal{P} .

Contact networks and transmission trees



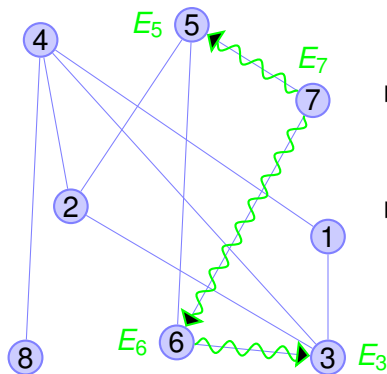
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Contact networks and transmission trees



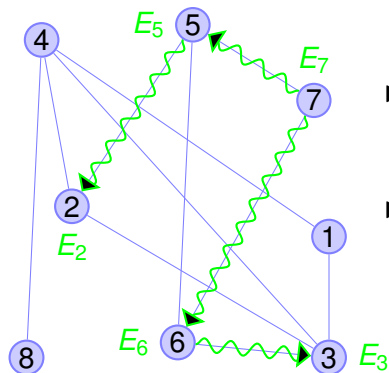
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Contact networks and transmission trees



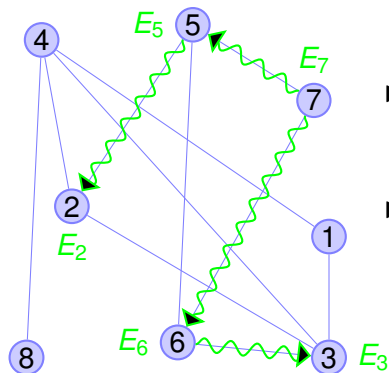
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Contact networks and transmission trees



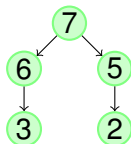
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Contact networks and transmission trees



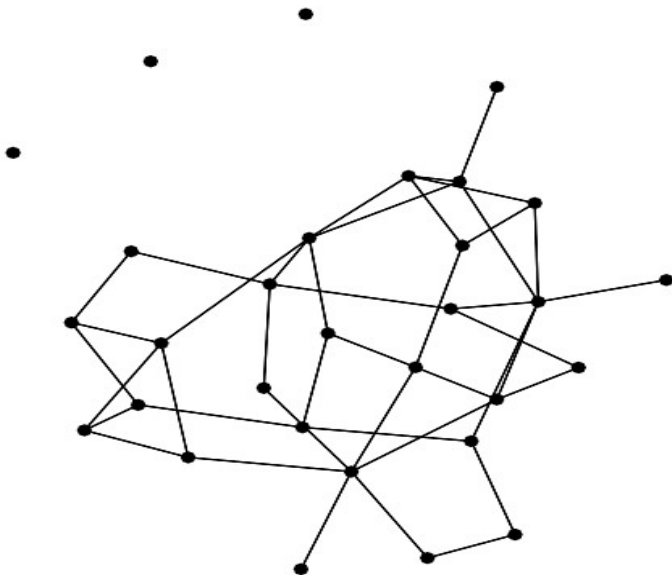
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Transmission tree, \mathcal{P} :



A network model: Bernoulli (Erdős-Rényi) network

For any nodes i and j , edge (i, j) exists with probability p .



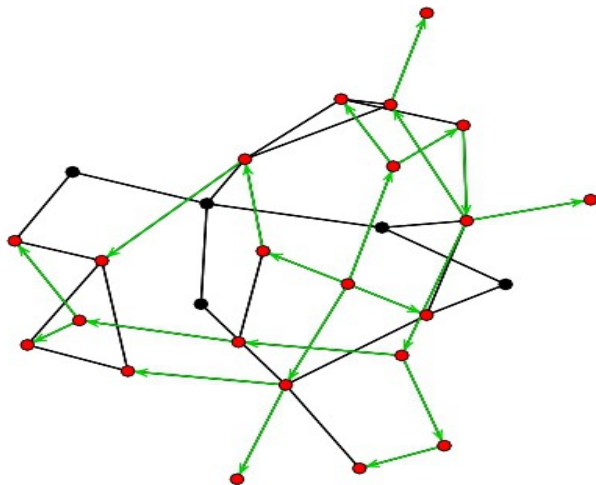
Modelling epidemics: Stochastic compartmental models

- ▶ The host population are considered to occupy different classes based on current disease status.
- ▶ The SEIR model has four classes:

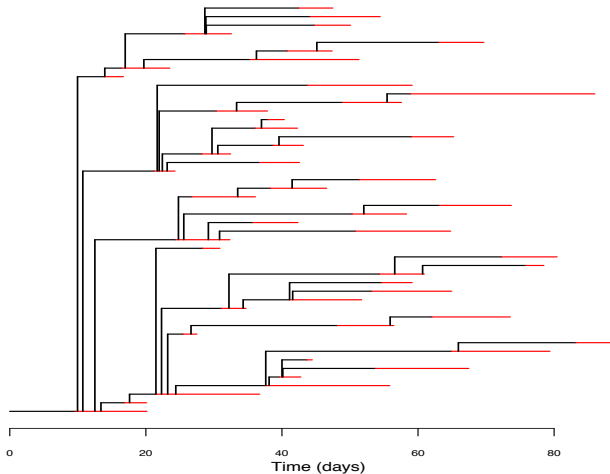
Susceptible \rightarrow Exposed \rightarrow Infectious \rightarrow Removed

- ▶ The $S \rightarrow E$ transitions have exponential waiting times, while $E \rightarrow I$ and $I \rightarrow R$ have gamma distributed waiting times.

Modelling epidemics: Stochastic compartmental models

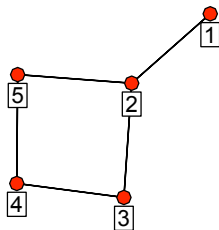


The path the epidemic follows describes a tree. The virus mutates down the tree and we collect genomes from the leaves.

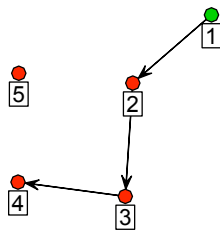


Can also use event time data where known

Node	Exposure Time	Infective Time	Removal Time
1	0.0	6.4	15.1
2	8.1	12.3	16.7
3	13.5	22.9	41.2
4	38.6	48.0	56.9



Contact Network



Transmission Tree

Parameters of interest

- ▶ Infection rate β
- ▶ Parameters θ and k of latent and recovery times
- ▶ η are the parameters of the network model
- ▶ Use Bayesian techniques to estimate posterior distributions for these parameters.
- ▶ To calculate likelihood, need also to estimate \mathcal{P} and \mathcal{G}

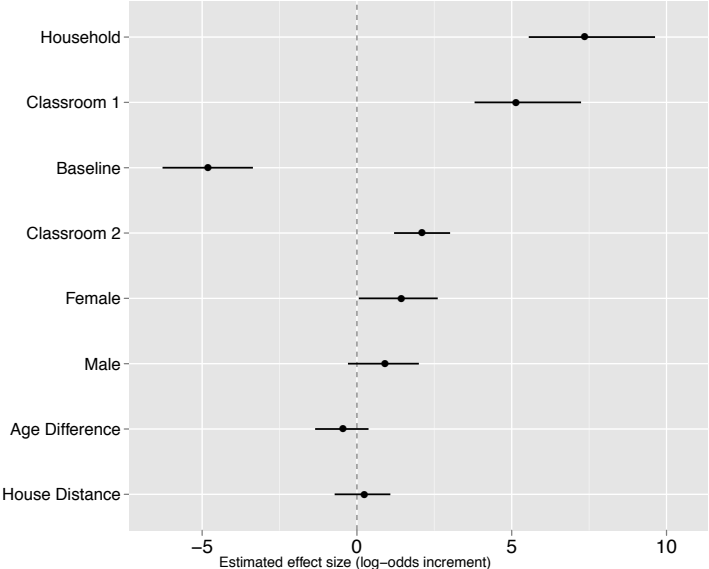
Hagelloch Measles Data

- ▶ We consider (unusually complete) data from a measles epidemic that spread through the small town of Hagelloch, Germany, in 1861.
- ▶ The data contain proxies for the Infective and Recovery times.
- ▶ We have to infer the Exposure times.

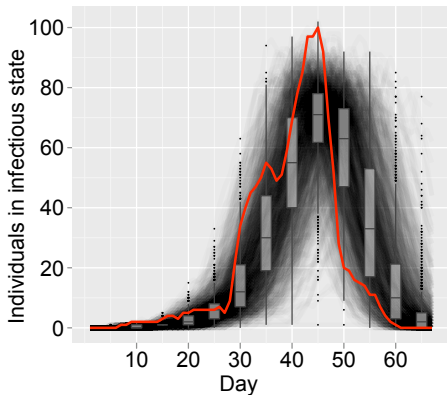
Node i	E_i	I_i	R_i
1	?	6.4	15.1
2	?	12.3	16.7
3	?	22.9	41.2
\vdots	\vdots	\vdots	\vdots

- ▶ Assume 188 individuals infected comprise the complete susceptible population so $N = 188$.

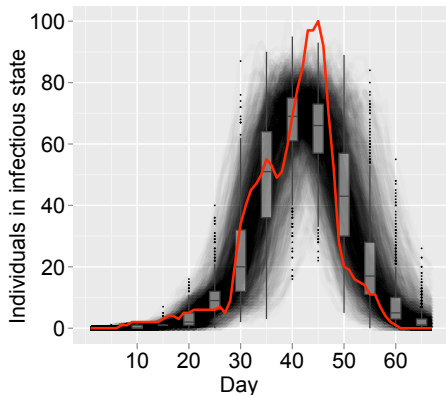
Posterior Estimates for η



Model checking — simulations (black) vs real (red)



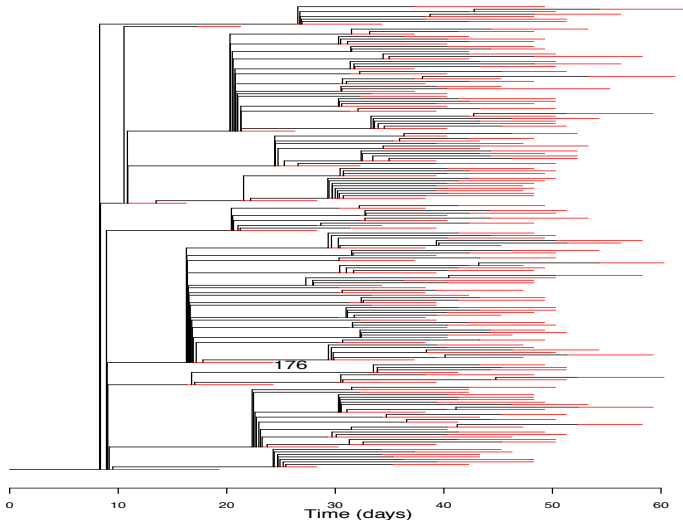
One edge probability



Multiple edge probabilities

Time data tell us little about who infected whom (tree topology)

An typical posterior sample of \mathcal{P}



Time data tell us little about who infected whom (tree topology)

