Evolutionary epidemiology in the 21st century

Data integration and modelling strategies

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Acknowledgements



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Who's this guy?

- Have a degree in Biology, can't tell an insect from a spider;
- PhD student in Evolutionary Biology (!) at the University of Edinburgh;
- Work mainly in Quantitative Biology;
- Interests include Markov chain Monte Carlo, complex networks and risk modelling.

Evolutionary epidemiology

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Challenges and opportunities

Methodological issues, data collection and handling.

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Which areas of mathematics are more heavily involved.

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Much more work is needed

We should prepare for an era of plenty.

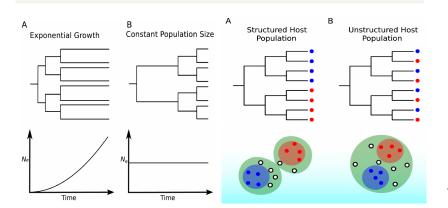
Motivation

Phylodynamics of fast-evolving viruses

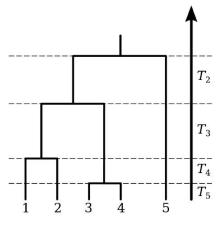
Inferring spatial and temporal dynamics from genomic data:

Phylogenies*!

* plus complicated models



Trees and the coalescent



Let T_n denote the time for n lineages to *coalesce*, i.e., merge into one ancestral lineage, in a population of size N_ℓ . Then:

$$\begin{split} Pr(T_n = t) &= \lambda_n e^{-\lambda_n t} \\ \lambda_n &= \binom{n}{2} \frac{1}{N_e} = \binom{n}{2} \frac{1}{N_e \tau} \end{split}$$

where N_ℓ is the effective population size and τ is the generation time. Let T_{mrca} denote the age of the most recent common ancestor:

$$\begin{split} \mathbb{E}[T_{\text{mrca}}] &= \mathbb{E}[T_n] + \mathbb{E}[T_{n-1}] + \dots + \mathbb{E}[T_2] \\ &= 1/\lambda_n + 1/\lambda_{n-1} + \dots + 1/\lambda_2 \\ &= 2N_e(1 - \frac{1}{n}) \end{split}$$

Figure: Figure 4 from Volz et al. (2013).

Data Integration I: Ebola epidemics in West Africa

[animation]

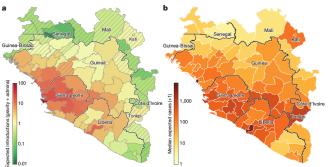
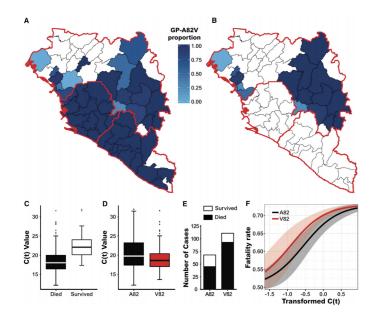


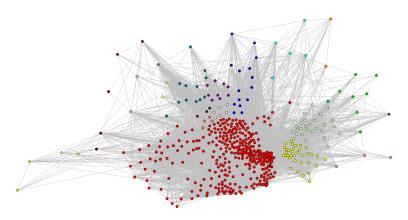
Figure 4 | Predicted destination consequences of viral dispersal number of EBOV imports into e 63 regions in Guinea, Sierra Leo Liberia (including 7 without rec in Guinea) and the surrounding of the neighbouring countries of Bissau, Senegal, Mali and Côte d expected number of EBOV expe locations in the phylogeographic imports to any location were cal the basis of the phylogeographic estimates and associated predict extended to apparently EVD-fre Supplementary Methods). b, Pre cluster sizes from the Bayesian C case data.

Data Integration II: GPA82V mutation and mortality



Potential applications

- Human mobility + case data = epidemiological models of spread and maintenance;
- Genomic data + environmental data = predictions of flow and case counts (e.g. Leptospirosis).



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Nature is complicated

We need better models to go along.

THE END