Antigenic flux in the influenza virus population

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The influenza virus



Gene flow in the influenza population





Approximately 1 in 20 sites change over the course of 10 years

Provides a chronological record of evolution

Influenza and measles differ in prevalence and age of infection



Unvaccinated cases Vaccinated cases Unknown vaccination status

Antigenic drift results in almost yearly influenza vaccine strain updates



Modeling antigenic phenotype

Influenza hemagglutination inhibition (HI) assay

Hemagglutination assay:



Without virus, red blood cell sink to bottom of well

With virus, cells form diffuse lattice

Hemagglutination inhibition assay:



Influenza hemagglutination inhibition (HI) assay



Without antibodies, agglutination of virus to RBC

Antibodies bind viruses, preventing agglutination



Developed by Derek Smith and colleagues

Uses multidimensional scaling (MDS) to position viruses in 2D space such that the distances in this space best fit the HI assay titres.



Smith et al. 2004. Mapping the antigenic and genetic evolution of the influenza virus. Science.

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Multidimensional scaling (MDS)

Maximum titer (in log2 space) of serum j

HI titer of virus *i* vs serum *j*

Observed HI distance between virus *i* and serum *j*

Map distance between virus *i* at X_i and serum *j* at Y_j

The goal of **MDS** is to find locations that minimize the difference between HI distance and map distance

 $d_{ij} = s_j - H_{ij}$

 H_{ij}

 s_j

 $\delta_{ij} = ||X_i - Y_j||_2$

 $\sum_{(i,j)\in\mathcal{I}} \left(\delta_{ij} - d_{ij}\right)^2$

Bayesian multidimensional scaling (BMDS)

In **Bayesian MDS**, this is probabilistically reformulated as

For example, given $\delta_{ij} = 2$ and $s_j = 8.3$

$$H_{ij} \sim \text{Normal}(s_j - \delta_{ij}, \varphi^2)$$



Oh and Raftery. 2001. Bayesian multidimensional scaling and choice of dimension. J Am Stat Assoc.

$p(\mathbf{X}, \mathbf{Y}|\mathbf{H}) \propto p(\mathbf{H}|\mathbf{X}, \mathbf{Y}, \varphi) p(\mathbf{X}, \mathbf{Y}, \varphi)$

Integration through Markov chain Monte Carlo (MCMC)



BEAST: Bayesian Evolutionary Analysis by Sampling Trees

Bayesian MDS results

Antigenic map of H3N2 influenza from 1968 to 2011



Dimen 1

Antigenic drift of H3N2 influenza



Integrating phylogeny and cartography

Phylogeny of H3N2 virus sequences



Modeling a continuous character as Brownian motion diffusion



Lemey et al. 2010. Phylogeography takes a relaxed random walk in continuous space and time. Mol Biol Evol.

$p(\mathbf{X}, \mathbf{Y}|\mathbf{H}) \propto p(\mathbf{H}|\mathbf{X}, \mathbf{Y}, \varphi) p(\mathbf{X}|\sigma, \tau) p(\mathbf{Y}, \varphi, \sigma, \tau)$

Comparison of MDS and diffusion models



Dimen 1

Comparison of MDS and diffusion models



Dimen 1

Antigenic drift driven by trunk lineage



Antigenic drift driven by trunk lineage



Linking genetic and phenotypic change

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 Bedford T, Suchard MA, Lemey P, Dudas G, Gregory V, Hay AJ, McCauley JW, Russell CA, Smith DJ, Rambaut A. 2013. Integrating influenza antigenic dynamics with molecular evolution. arXiv: 1304.3637.

Source code:

code.google.com/p/beast-mcmc/