



# What can we learn from Whole-Genome-Sequencing?

Uncovering transmission patterns using sequence data and phylodynamics

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- What phenotype/properties does the pathogen have ?
  - Host range and transmissibility ? Drug resistance ?
- Where did it come from ?
  - Which host species
  - Which locations ?
  - Who infected whom ?
  - Co-infection and mixing ?

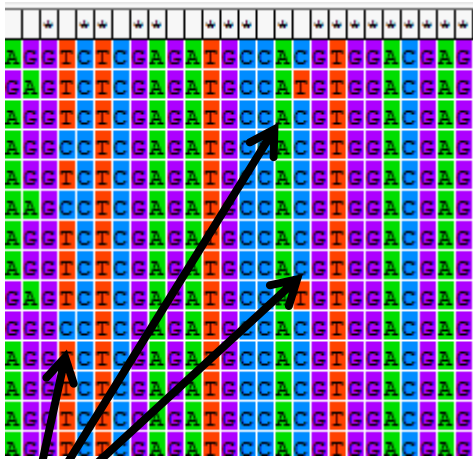
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t a g c g c a c g g a g a t g a g a g c c c g c g c a a c t a a g a c g c c a g a c c c t c c  
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c c a .....  
... t a .....  
... t g .....  
... t g g c .....  
... a g a .....  
... a g a g c .....  
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... a c t . t .....  
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... a c t . t t a t . g . . . t t g t g t t .....  
... a c t . t t a t . g . . . t . . . t c .....  
..... t ..
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Fine scale properties of genome  
Every SNP can be useful

# Pathogen Sequence Data

- Pathogen sequence data provides richer information than strain type
- Sequences accumulate mutations over time – classic picture (influenza)

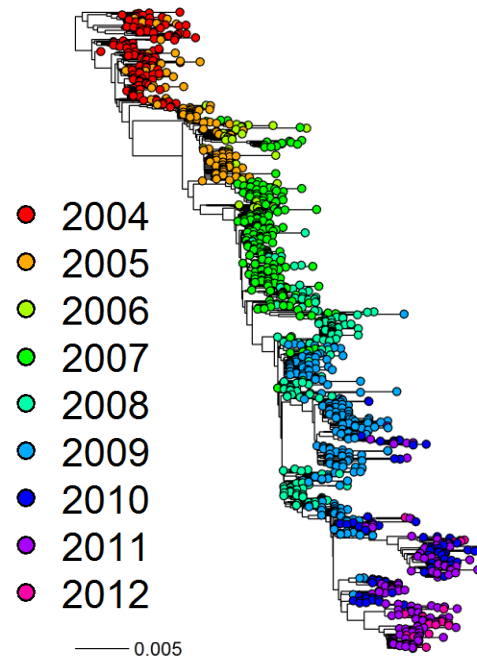
Sequences, one per row



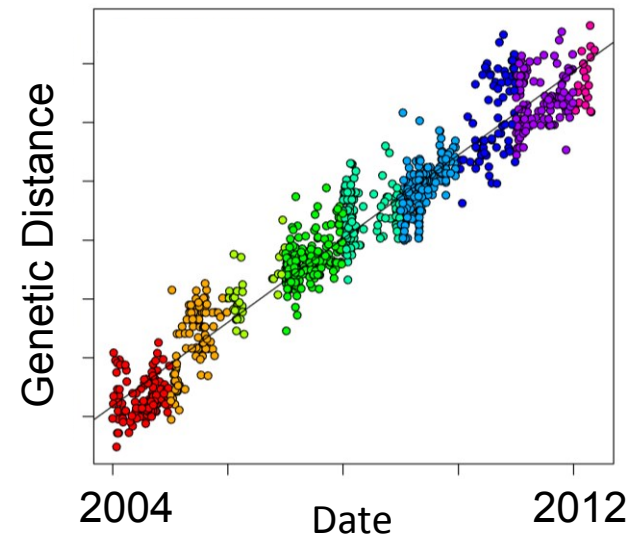
Mutations

Converted

Tree of Human Influenza



Genetic Distance from Root





How much sequence variation ?



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# Evolutionary Rates

	RNA Viruses	DNA Viruses	Bacteria
Replication & Evolution	Fast and error prone	Slower, more conserved	Slow
Genome size	8-14kb	20-200kb	4Mb
Mutations per year	10-100	1-20	0-1 ?

Classical Swine Fever  
 Bovine Viral Diarrhoea  
 Foot-and-Mouth

Segmented ssRNA

Avian influenza  
 Schmallenberg

African Swine  
 Fever

Segmented dsRNA

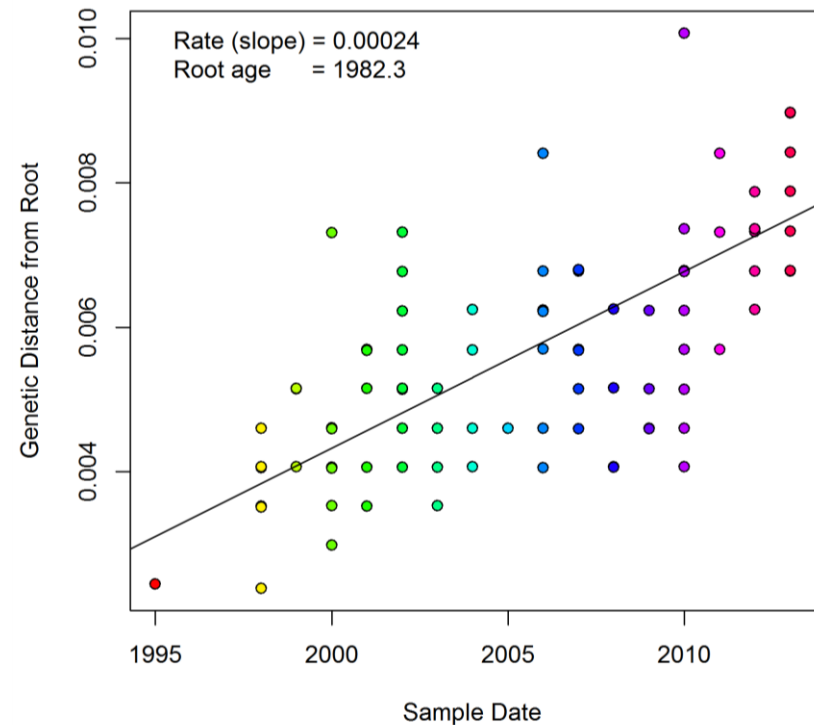
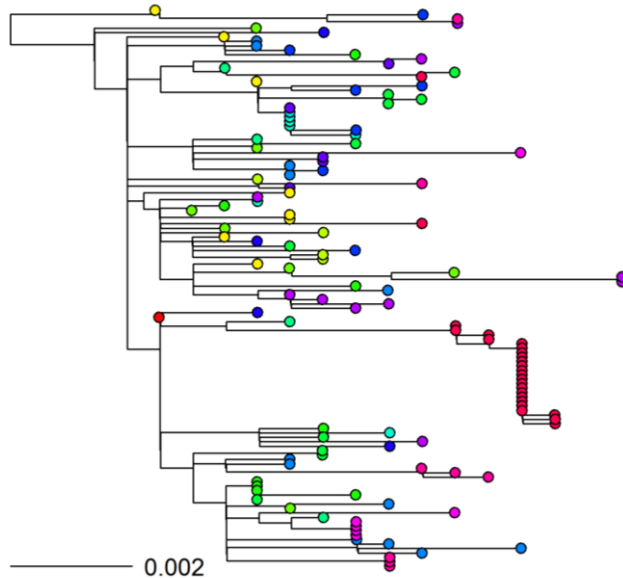
Blue Tongue  
 African Horse Sickness

Bovine Tb

# Mbovis example – one clade one region



- True rate estimates  $\sim 0.1$  SNP per lineage per year
- Example:
  - 126 samples, date range = 18 years
  - Number of variable sites = 309 (concat. SNPs)





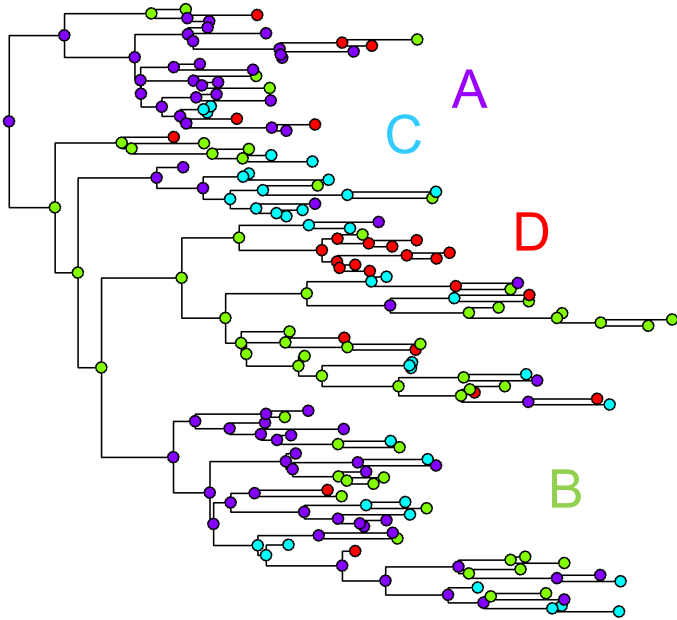
# Inferring Transmission Patterns



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## Tree with Location Traits



## Transition Rate Matrix (M)

	A	B	C	D
A	-	B→A	C→A	D→A
B	A→B	-	C→B	D→B
C	A→C	B→C	-	D→C
D	A→D	B→D	C→D	-

Probability of Ancestral state ( $x'$ ), given branch length  $t$  and child state  $x$ :

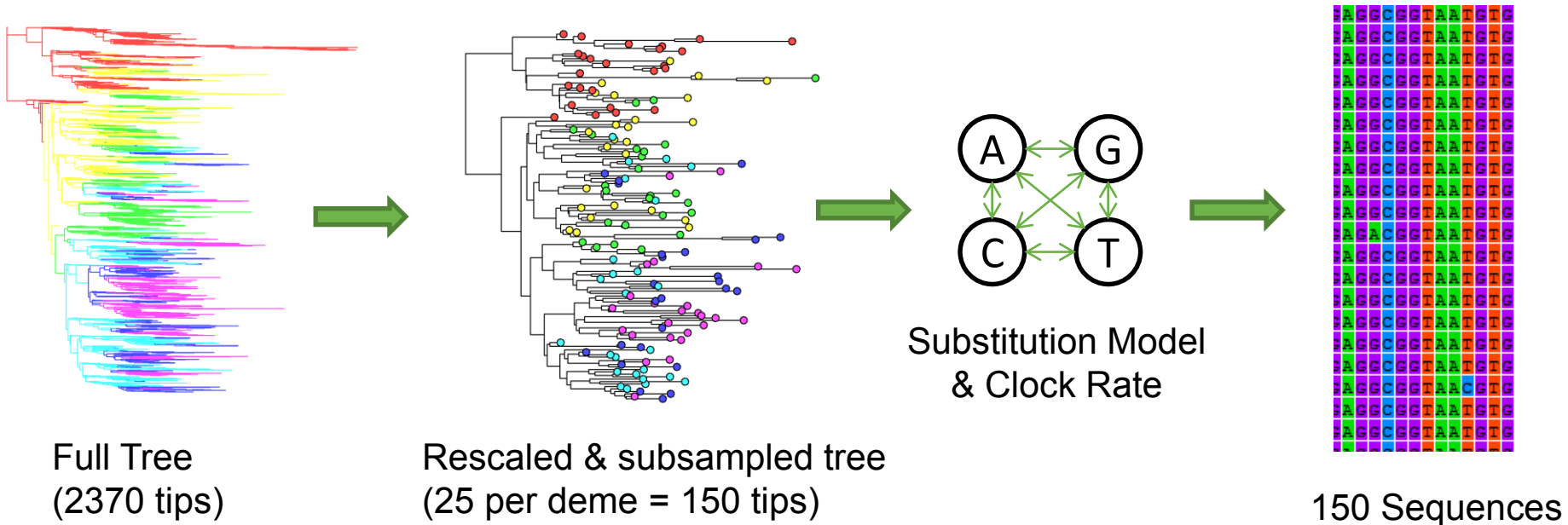
$$p(x'|t) \sim e^{Mt} x$$

- Add locations to phylogenetic tree
- Estimate transition rates between locations along branches
- Transmission pattern represented by rate matrix



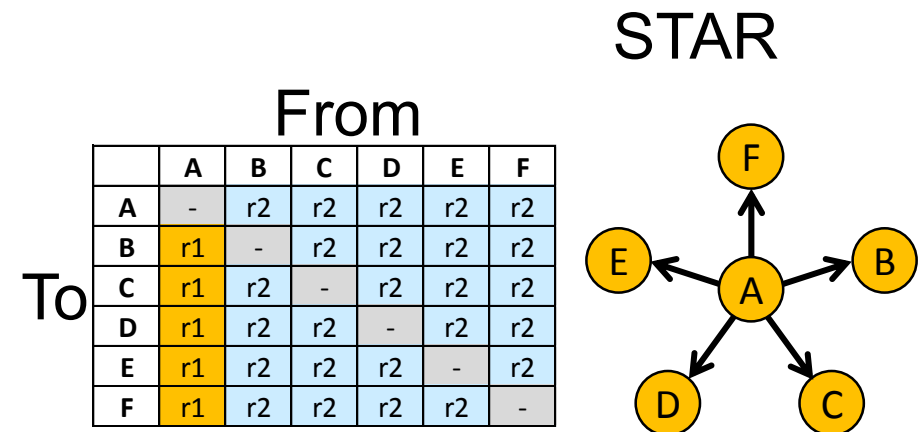
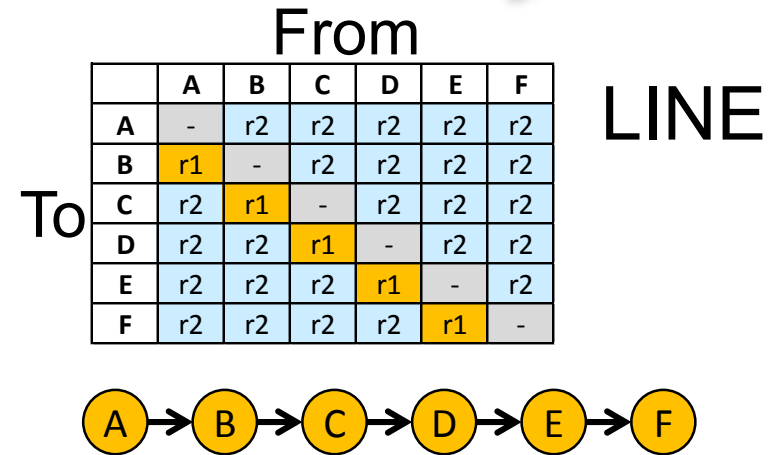
# Simulate Sequences

- Examine phylodynamic situations by simulation
  - output true transmission tree and phylogenetic tree
- Simulate sequences down tree
  - Use different mutation rates and lengths, equivalent to:
    - 0.025 – 2 substitutions per genome per year
    - samples spanning 15 years
- Total number of SNPs in data: 10 - 750



# Simulate Trees

- DiscreteSpatialPhyloSimulator (DSPS) to simulate infection over structured population
- Individual based model, individuals are farms
- 6 regions (random mixing within demes)
  - 500 farms per deme
  - Each farm is SIR; beta = 0.1, gamma = 0.05
  - Infection between demes = 0.1
  - Demes connect in LINE or STAR network



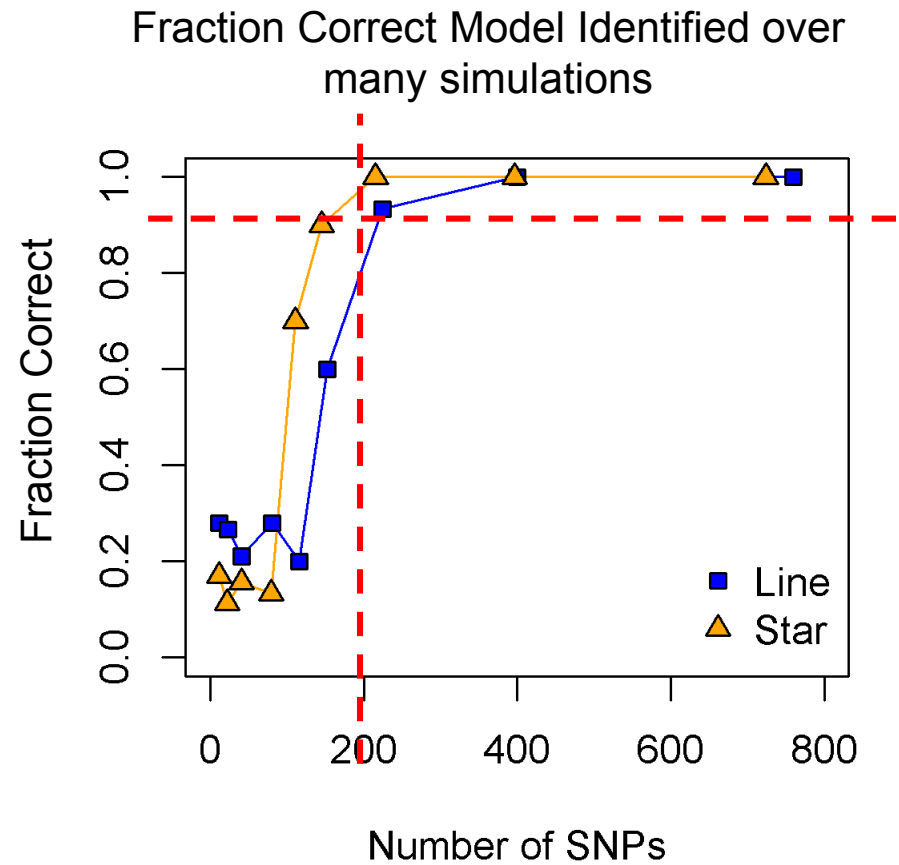
Code available – currently tidying & validating...

<https://github.com/hxnx-sam/DiscreteSpatialPhyloSimulator>

# Finding the correct pattern



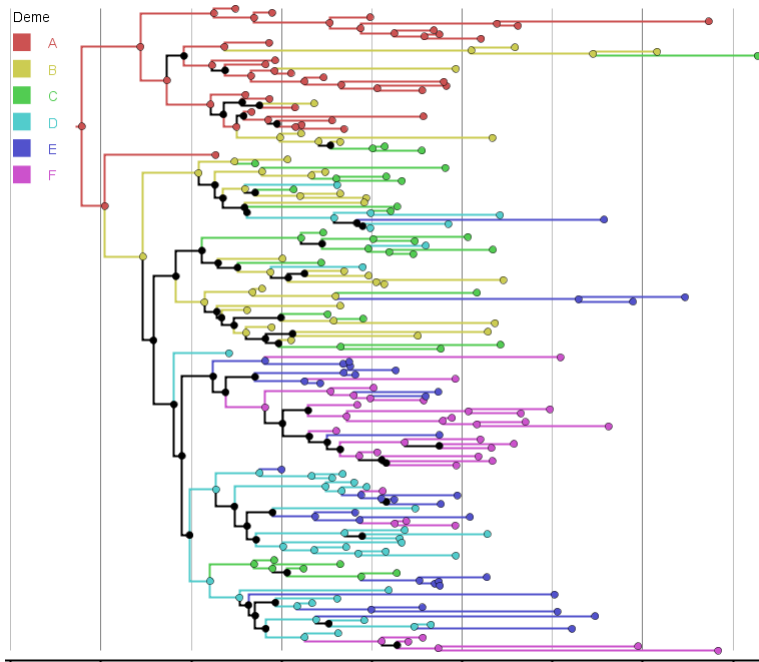
- Reconstruct trees from simulated sequences using Neighbour Joining
- Calculate likelihood of line, reverse line, star, reverse star models upon reconstructed trees
- Fraction of simulations where correct transmission pattern found is a function of number of mutations



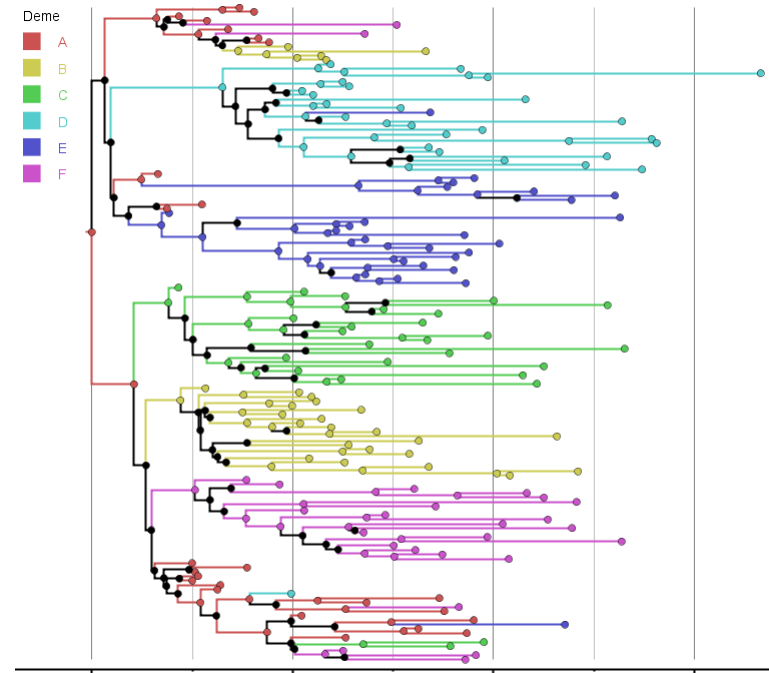
200 SNPs => 90% correct  
(~0.5 mutations per year)

# Phylodynamics using BEAST

- Infer trees and transition rate matrix with BEAST
- Use Line and Star scenarios with differing sequence lengths and mutation rates (slow & short, moderate, fast & long)



LINE



STAR

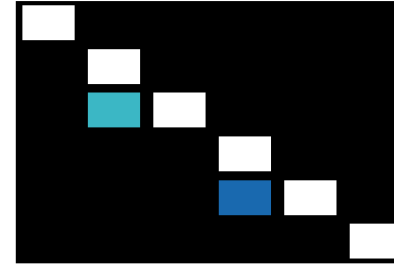
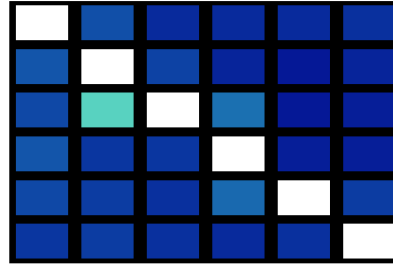
# Detecting Line Population Structure



## Full Rate Matrix

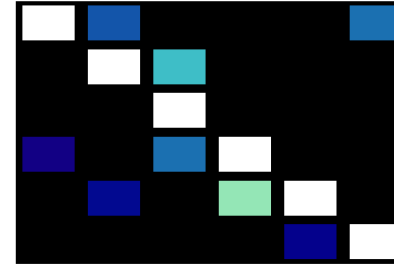
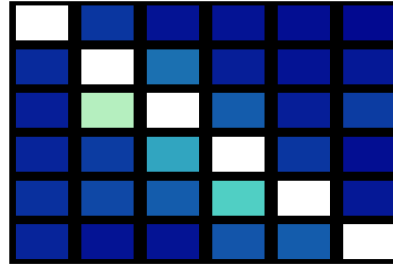
## Significant Rates

(i) Short and slow  
22 SNPs



LINE  
Population  
Structure

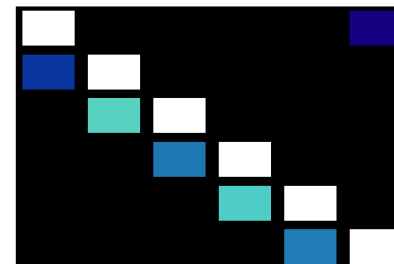
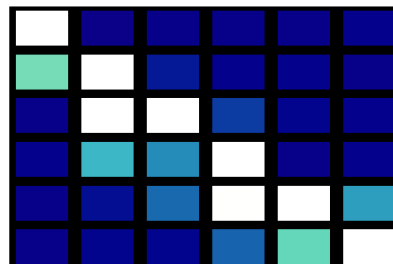
(ii) Moderate  
80 SNPs



BVD single region  
(3 years)

**TB WGS**  
(10-20 years)

(iii) Long and fast  
400 SNPs



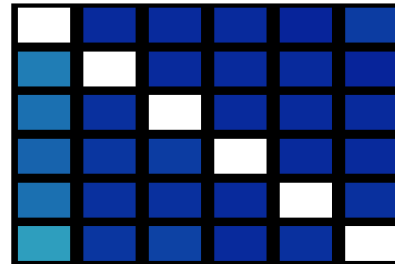
Flu segment  
(3 years)

# Detecting Star Population Structure

## Full Rate Matrix

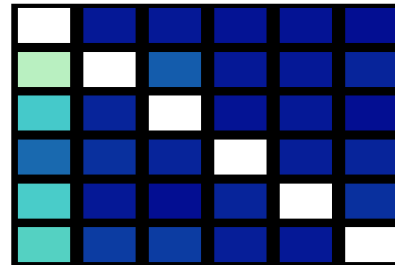
## Significant Rates

(i) Short and slow  
22 SNPs



STAR  
Population  
Structure

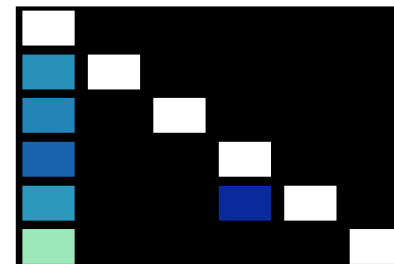
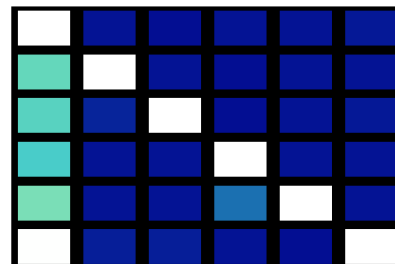
(ii) Moderate  
80 SNPs



BVD single region  
(3 years)

**TB WGS**  
(10-20 years)

(iii) Long and fast  
400 SNPs



Flu segment  
(3 years)



Example

With multiple traits !



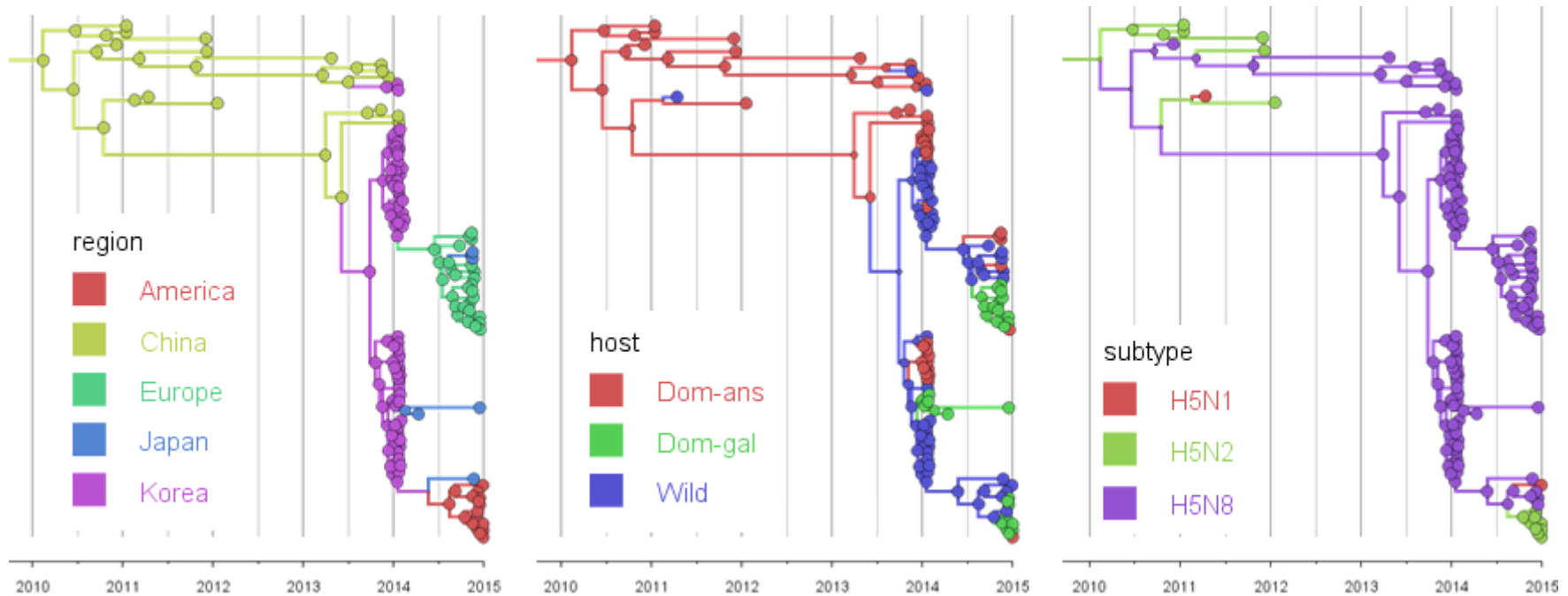
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# Phylodynamics with multiple traits



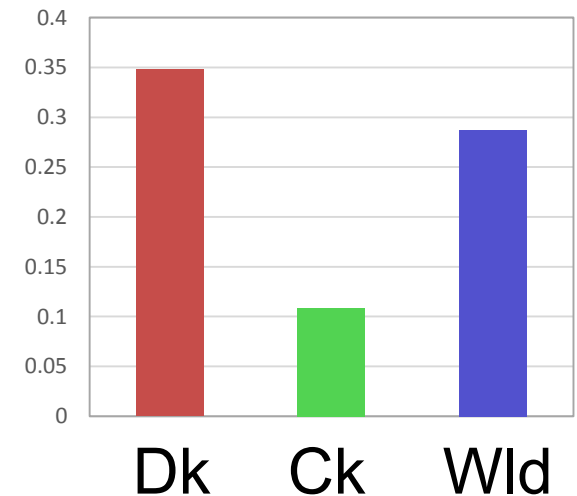
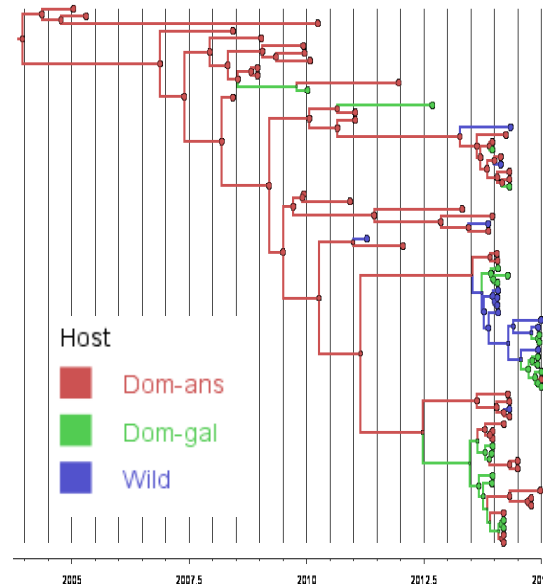
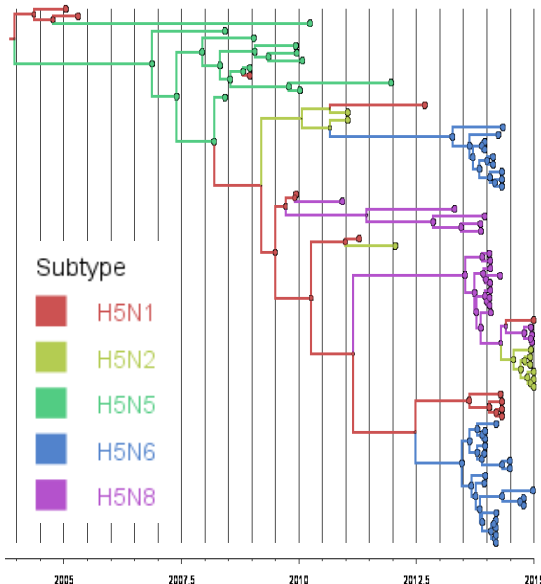
- Recent HPAI Avian influenza in UK and North America
- Where did it come from ?
- Generate time resolved trees from HA sequences
- Include region, host and subtype as discrete traits





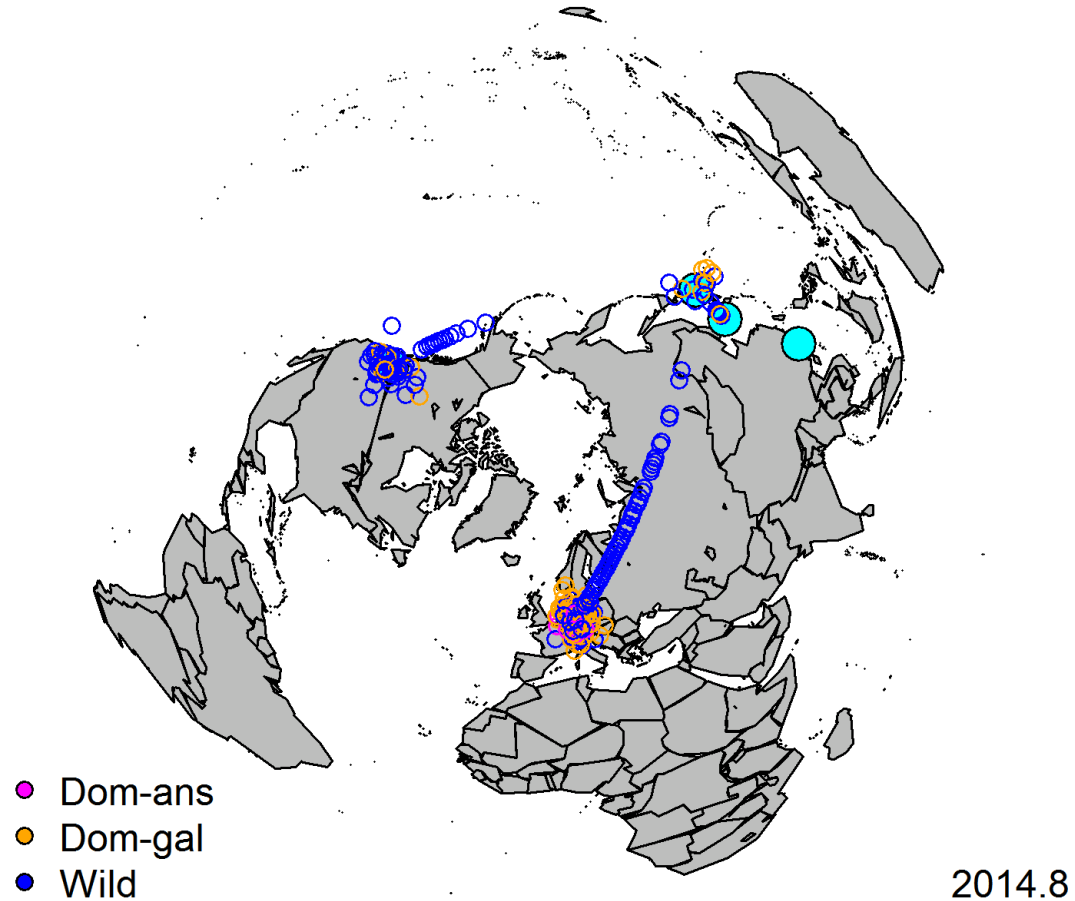
# Correlation of traits

- Map host and subtype on same set of trees
- Count subtype changes on duck, chicken or wild birds only branches
- Find more reassortment in ducks and wild birds (anseriformes)



# Phylodynamics with host and spatial information

## Shows dispersion by one host species



(one image of movie)



# Distinguishing transmission patterns



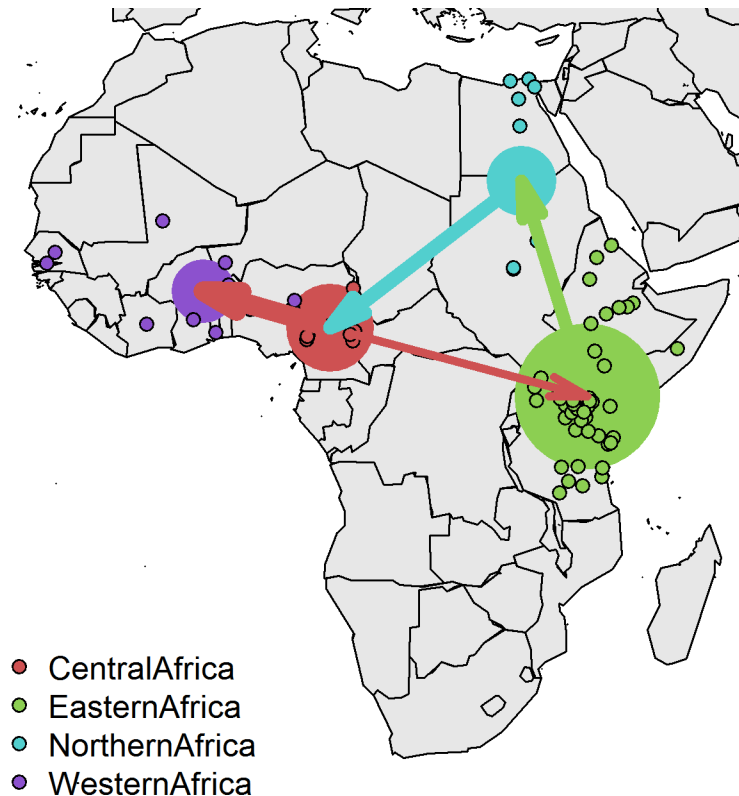
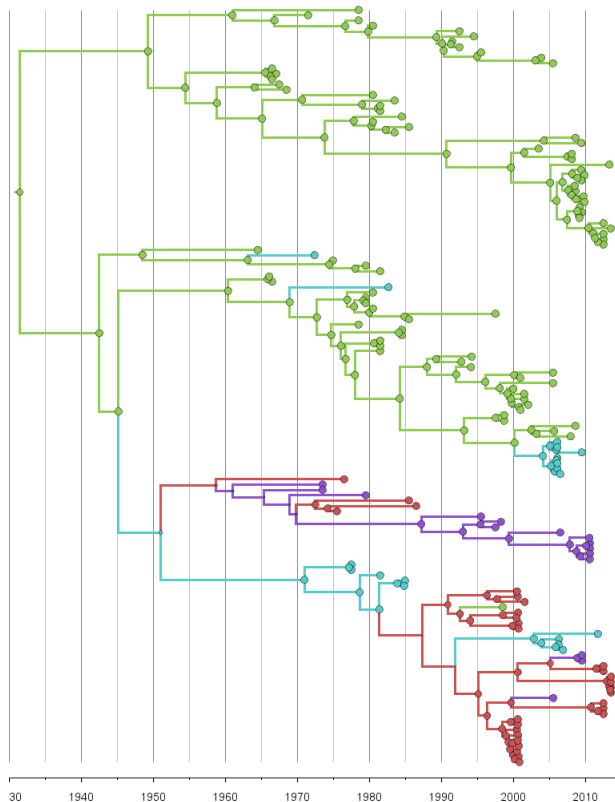
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- Are the transmission patterns due to known animal movements ?
- Or is there something else ?
- Method 1:
  - Infer rate matrix from discrete locations and calculate significant links between places using BSSVS
- Method 2:
  - use Latitude & Longitude and infer routes taken
- (Both) Compare to known movements (manually)

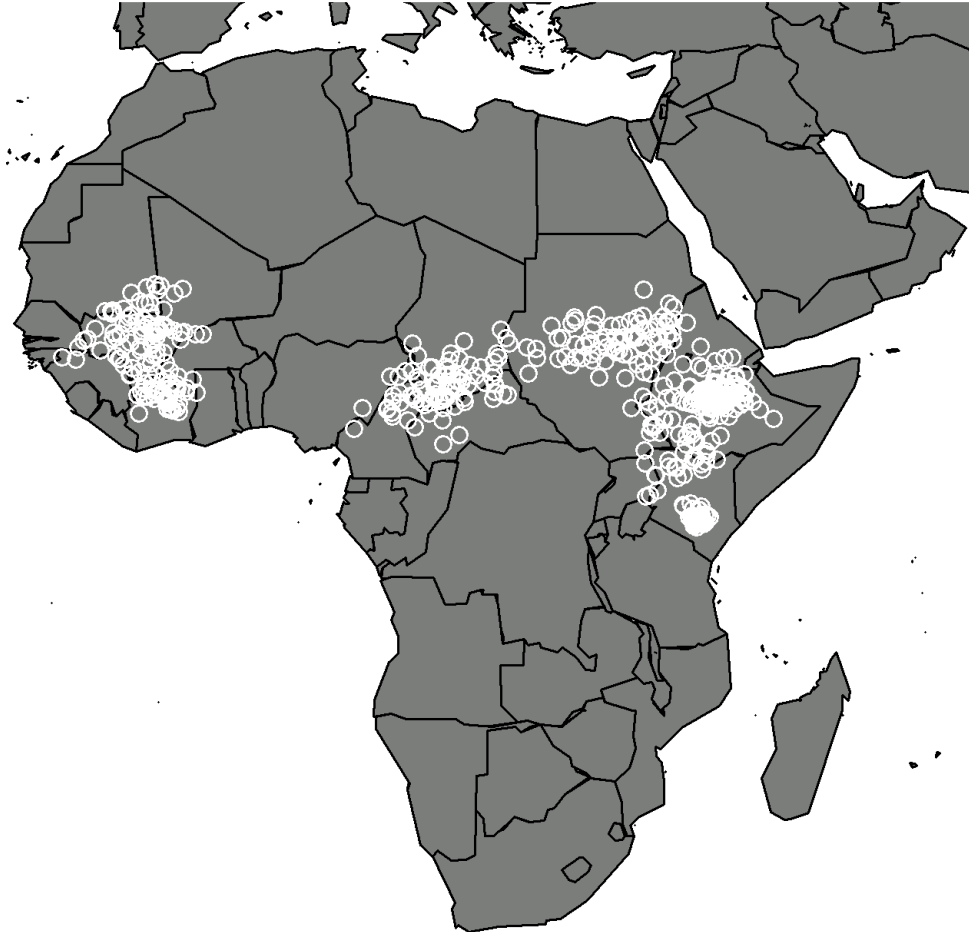
# FMD – Serotype A in Africa

- Sequences ~600 bases long of VP1
- 444 SNPs for 142 sequences in time scale 1964 – 2013 (49 years)
- Using regional groupings => 4 discrete states



Dispersion in space and time

Using Latitude & Longitude as continuous trait (lower clade)



1993 A

(one image of movie)

- Problems with Methods 1 & 2
  - Too many rates not enough different transmission events ?
  - Distances too far / diffusion not working ?
  - Why is it those rates or diffusion c/e anyway ?
- Use a Generalised Linear Model to parameterise the rates:

Transition Rate Matrix

For  $i=1$  to  $n$  predictors

Predictor Rate Matrix

$$\log \Lambda = \sum_i \delta_i \beta_i \log R_i$$

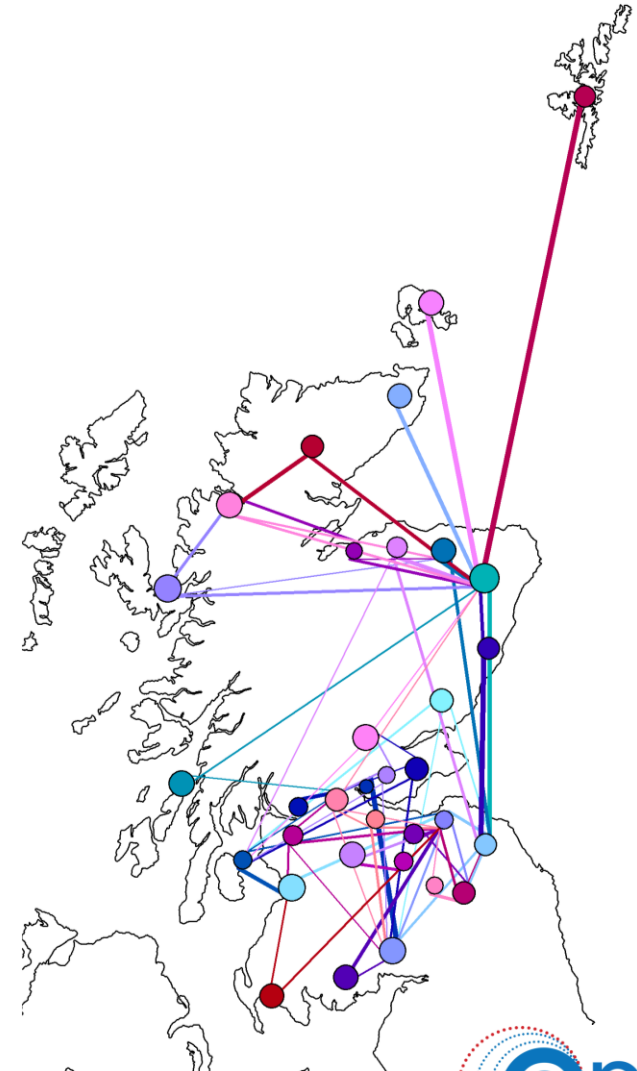
(for each MCMC step, propose  $\beta_i$ 's and  $\delta_i$ 's)

Indicator variable, delta dirac (0 or 1)

Coefficient of predictor matrix

**Now estimate the  $\delta$  and  $\beta$  instead of each rate matrix element**

- Simulate infection over population using “DiscreteSpatialPhyloSimulator”
- Simulate individual farms within 33 counties in Scotland
- Probability of infection between counties proportional to averaged movements of Cattle Tracing System
- Generate who-infected-who, but subsample to 10 sequences per county (massively undersampled!)
- Simulate sequences as before





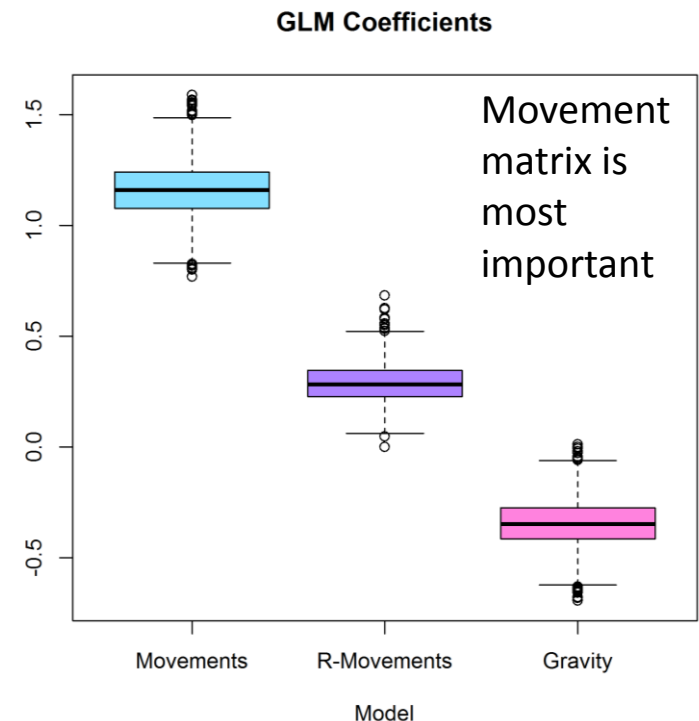
# Detecting Transmission Patterns



- Infer tree with Discrete traits model in BEAST
  - Model is 33 x 33 matrix (1056 rates)
  - Far too many individual rates !

- Use Generalised Linear Model
  - Predictor 1: Movement matrix
  - Predictor 2: Reversed movement
  - Predictor 3: Gravity Model

$$\frac{\text{Source size} \times \text{Dest. size}}{\text{distance}^2}$$



- Can distinguish between possible transmission patterns in principle

# Summary

- **Transmission pattern inference possible with WGS**
- **Distinguish between different spatial patterns**
- **Can find host species specific patterns**

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Thank you !



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