

# Analysing within- and between-host HIV genetic diversity with `phyloscanner` for detection of transmission and more

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# The BEEHIVE Project: Bridging the Epidemiology and Evolution of HIV in Europe

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## **With thanks to**

Nick Croucher  
Katrina Lythgoe  
Oliver Ratmann



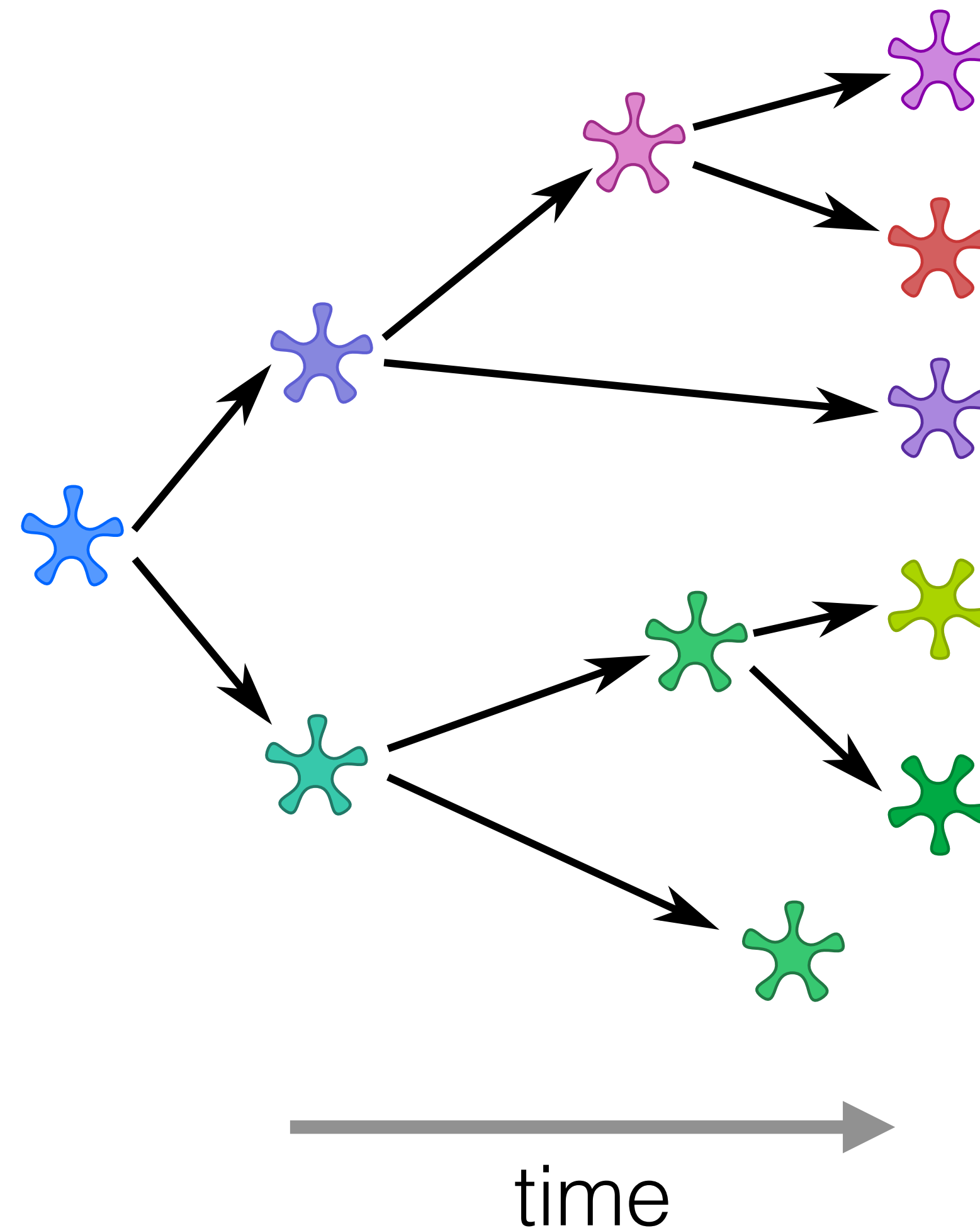


Problem: how can we target our interventions in an epidemic to have a greater impact?

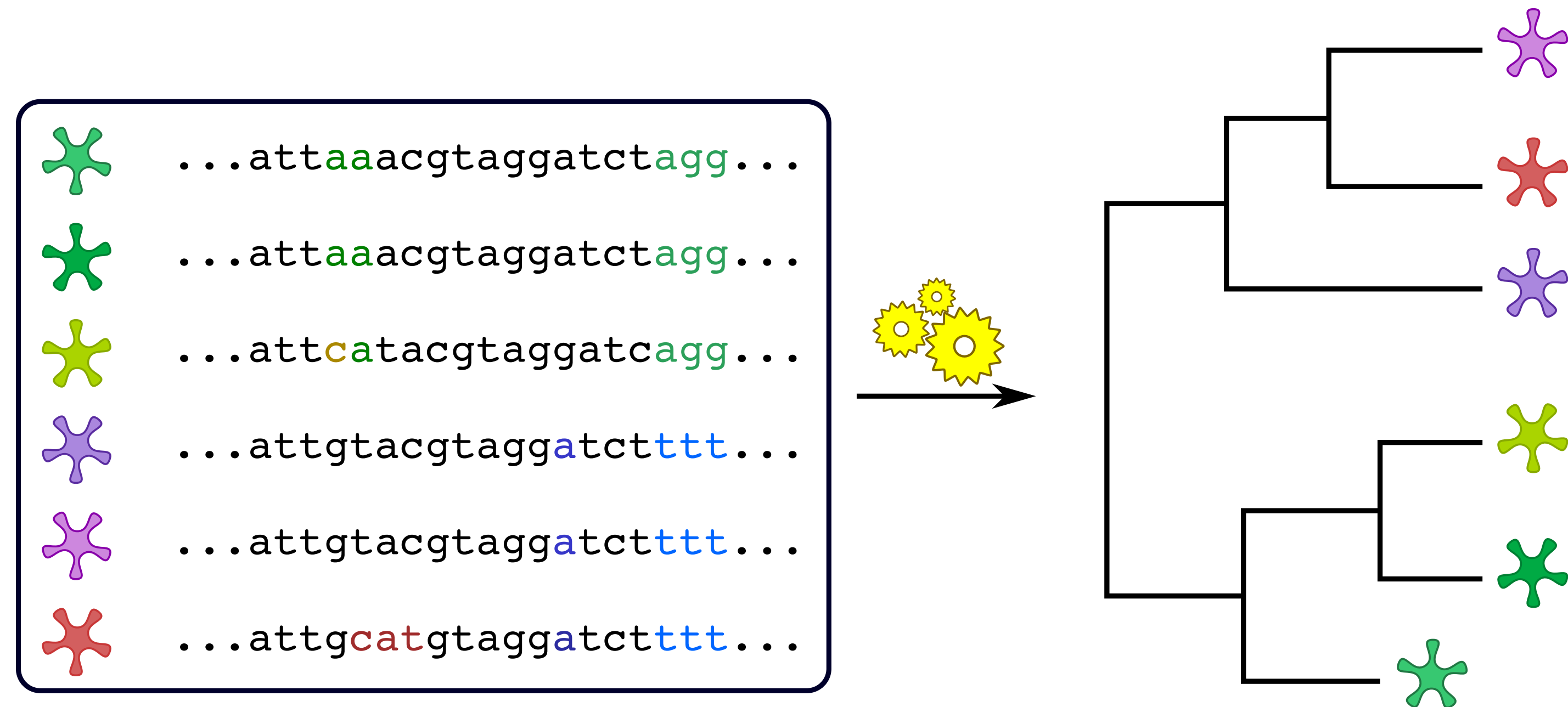
A solution: learn more about the epidemic through analysis of pathogen sequence data.

# Genetic changes (substitution) accumulate over time

**Substitution:** replacement of a nucleotide (e.g.  $a \rightarrow t$ )



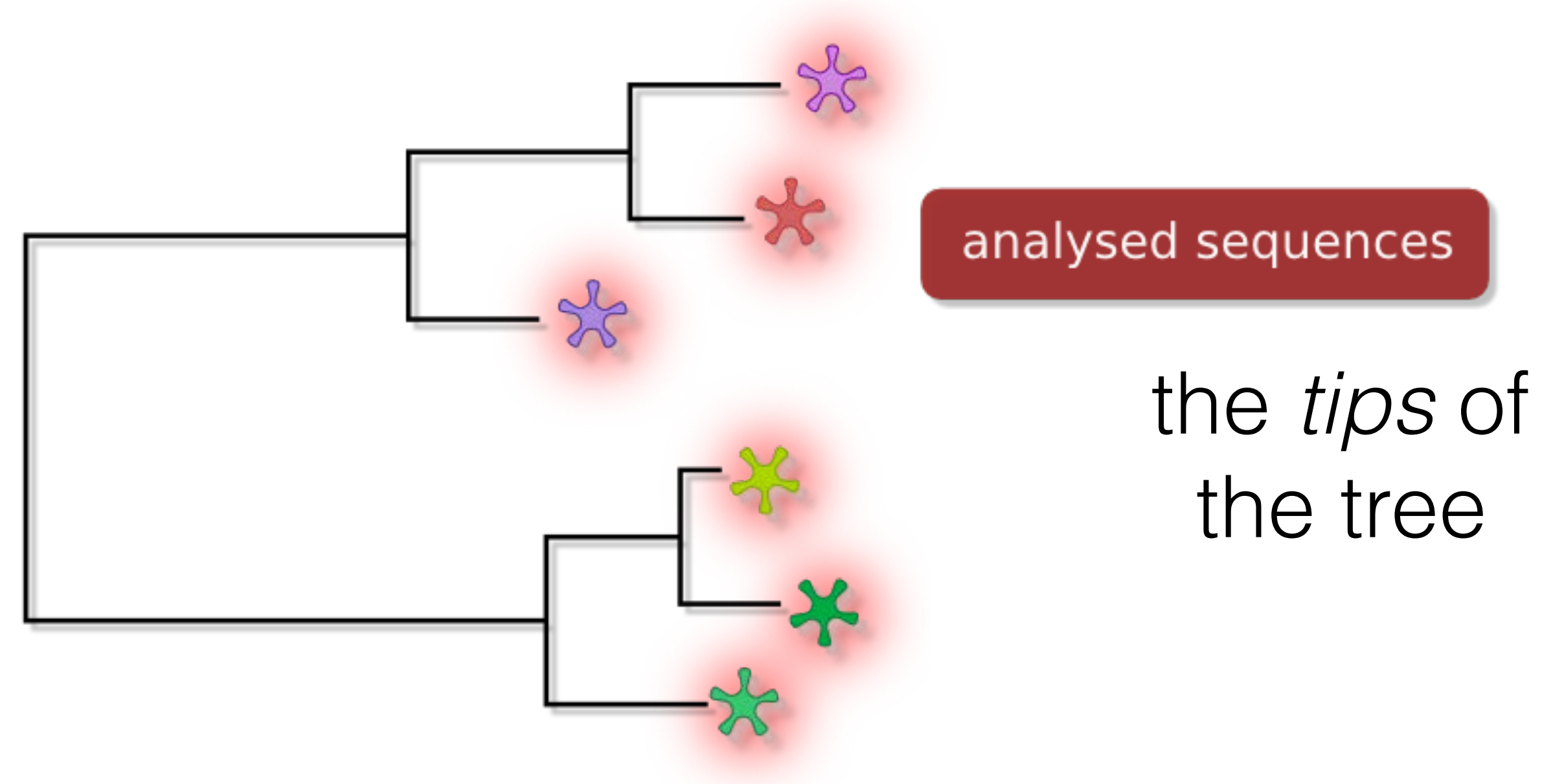
# Using substitution patterns to reconstruct the evolutionary history



Phylogenetics aim to reconstruct evolutionary trees (*phylogenies*) from genetic sequence data.

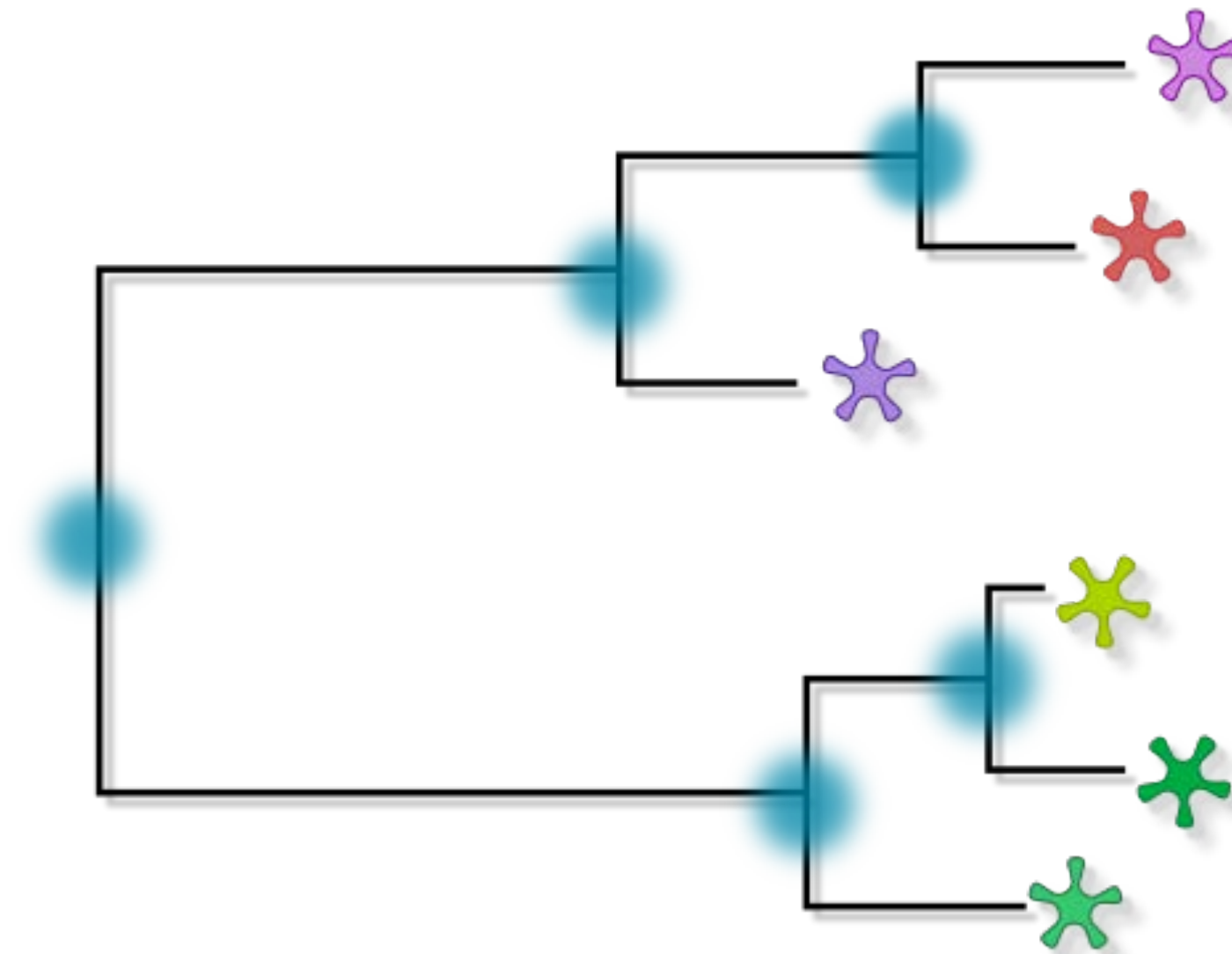


# Using trees to represent the evolutionary history



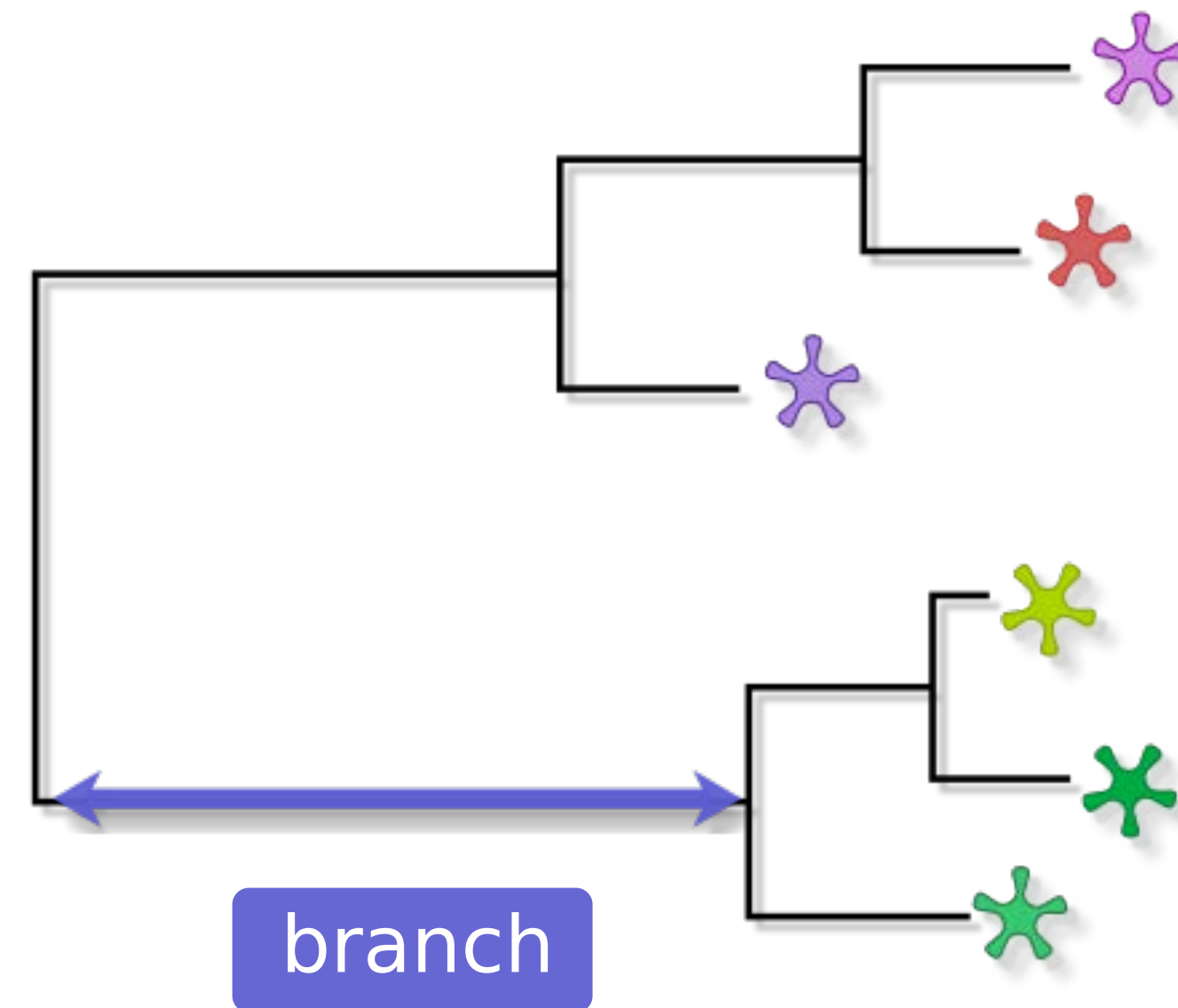
# Using trees to represent the evolutionary history

Most Recent Common Ancestors  
(MRCA)



the *nodes*  
of the tree

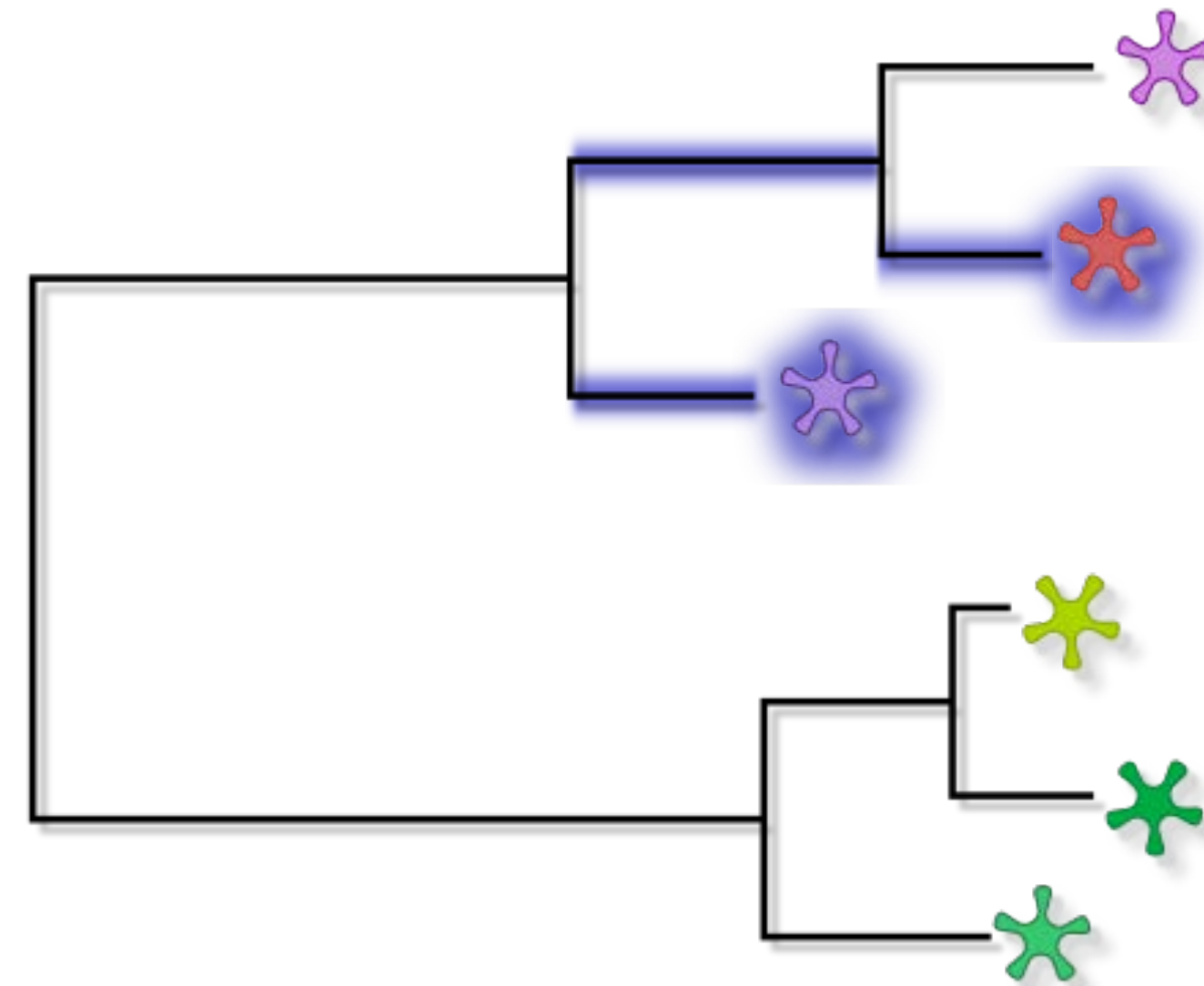
# Using trees to represent the evolutionary history



length = amount of evolution (**not time**, as a rule)



## Using trees to represent the evolutionary history



distances between tips

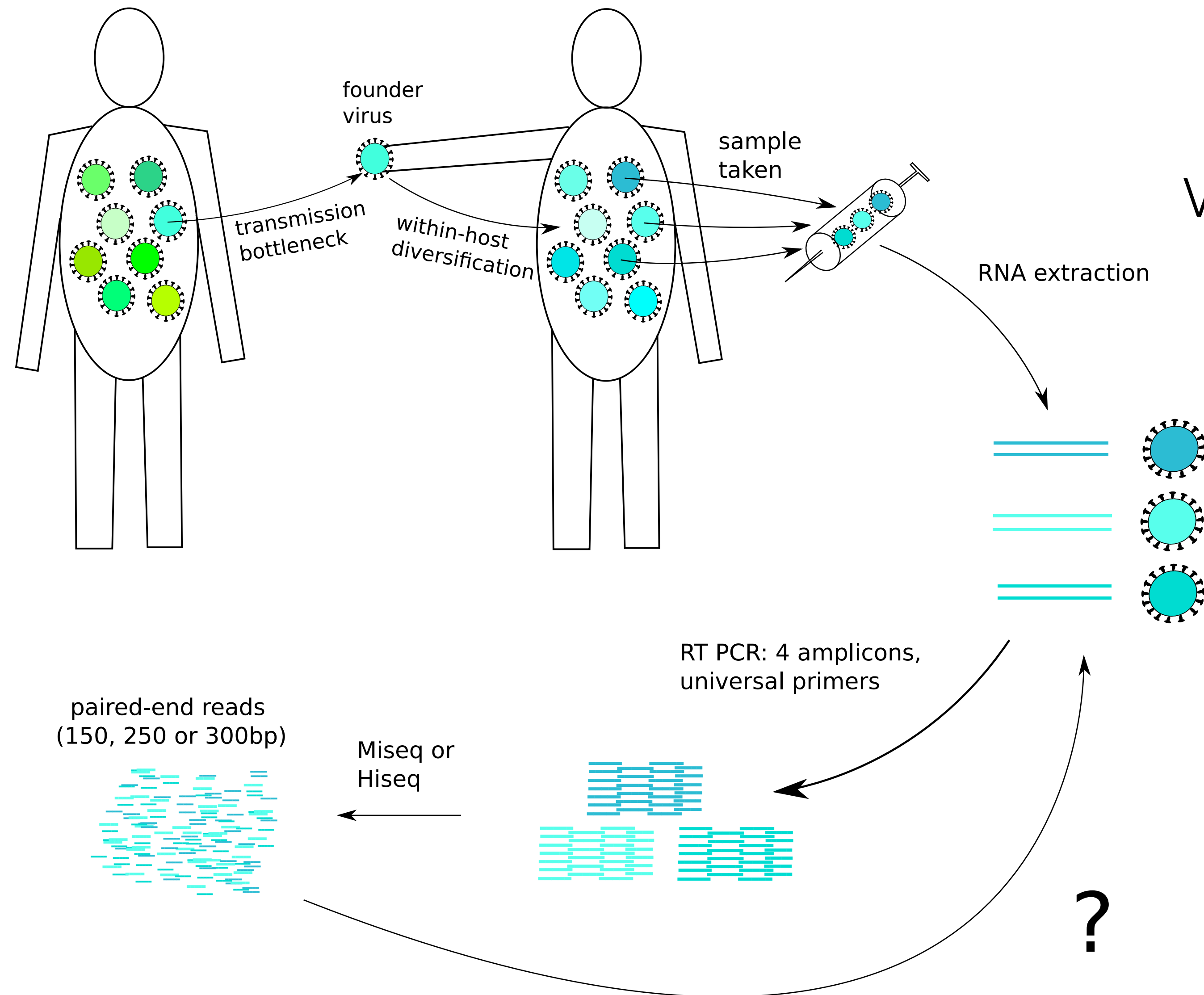
"*patristic*" distance: sum of branch lengths

Problem: to meaningfully interpret small differences between closely related viruses, we want *accurate* genomes.

Problem: to make robust statements, need *many* genomes.

Solution: high-throughput sequencing + accurate, scalable sequence processing.

# Our HIV Data



Cornelissen *et al.*  
Virus Research 2016

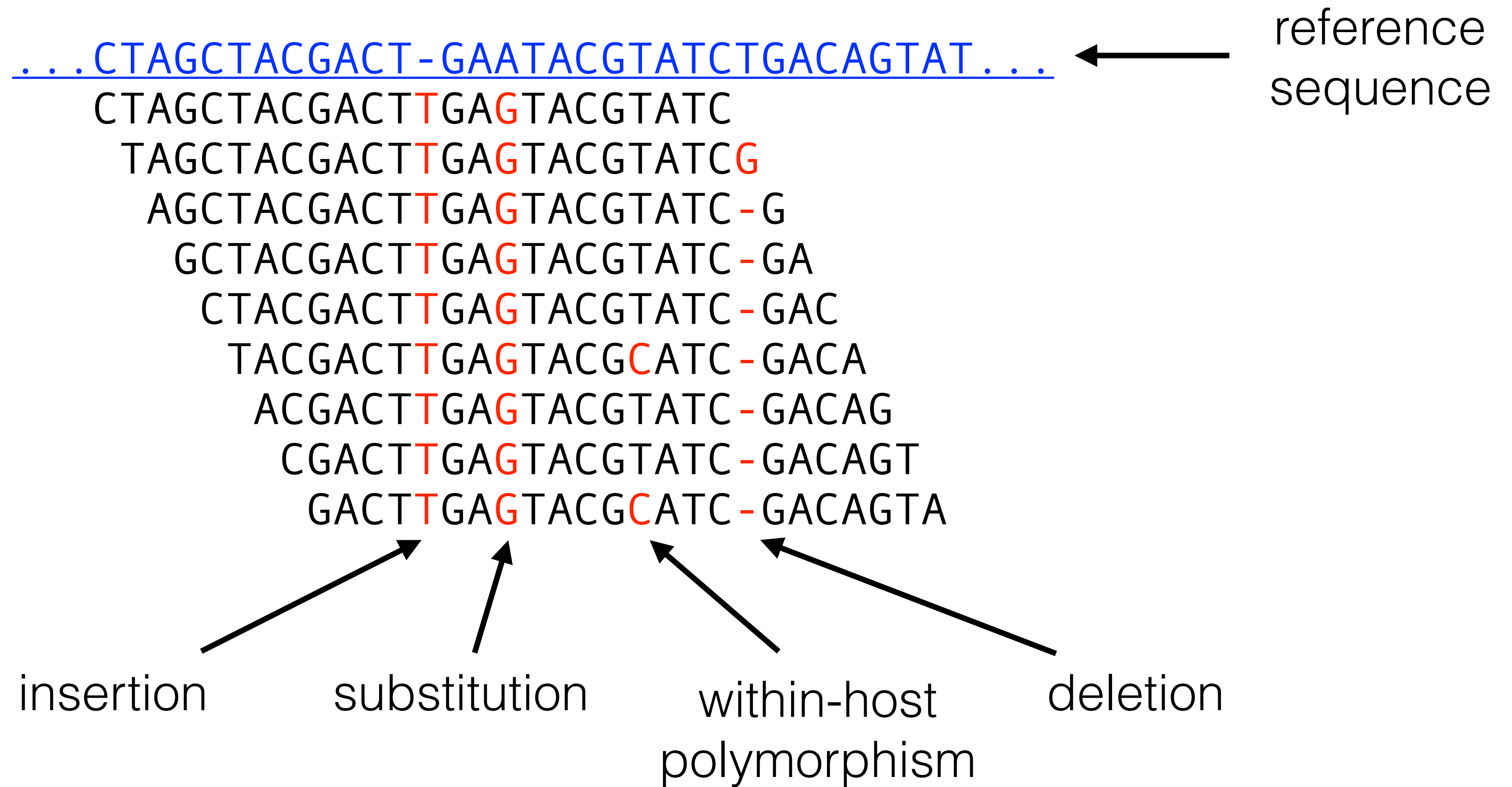
Gall *et al.*  
J. Clin. Microbiol.  
2012



↑ visiting, not helping



# Mapping Reads to a Reference

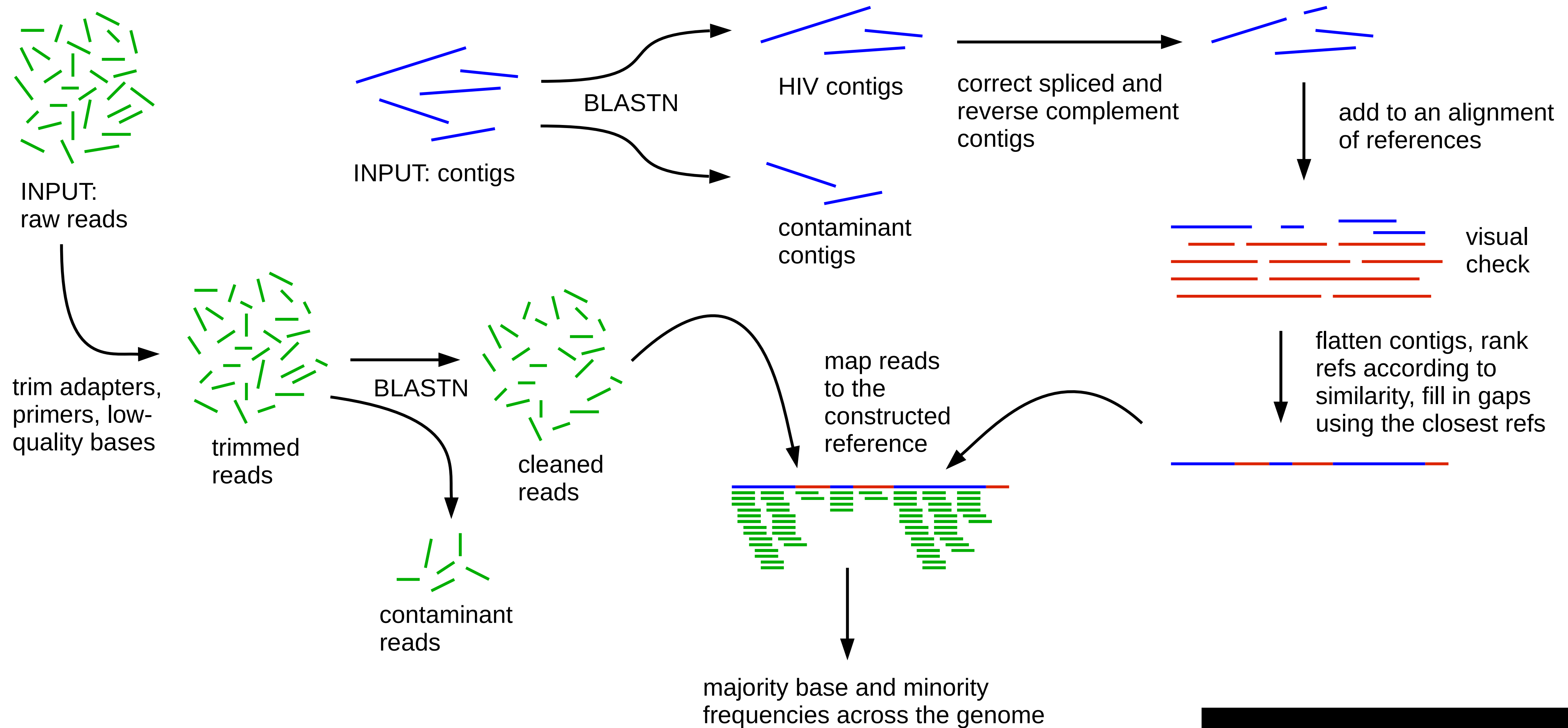


However, the more different a read is from its reference, the more likely it is to be aligned incorrectly or not at all.  
So **mapping causes biased loss of information.**

e.g. use the same reference for mapping for a group of individuals: bias their viruses to look similar to each other  
→ false inference of transmission.

e.g. use old references for mapping new samples  
→ false inference of slow evolution

# shiver - Sequences from HIV Easily Reconstructed



Wymant *et al.*, Virus Evolution 2018

[github.com/ChrisHIV/shiver](https://github.com/ChrisHIV/shiver)

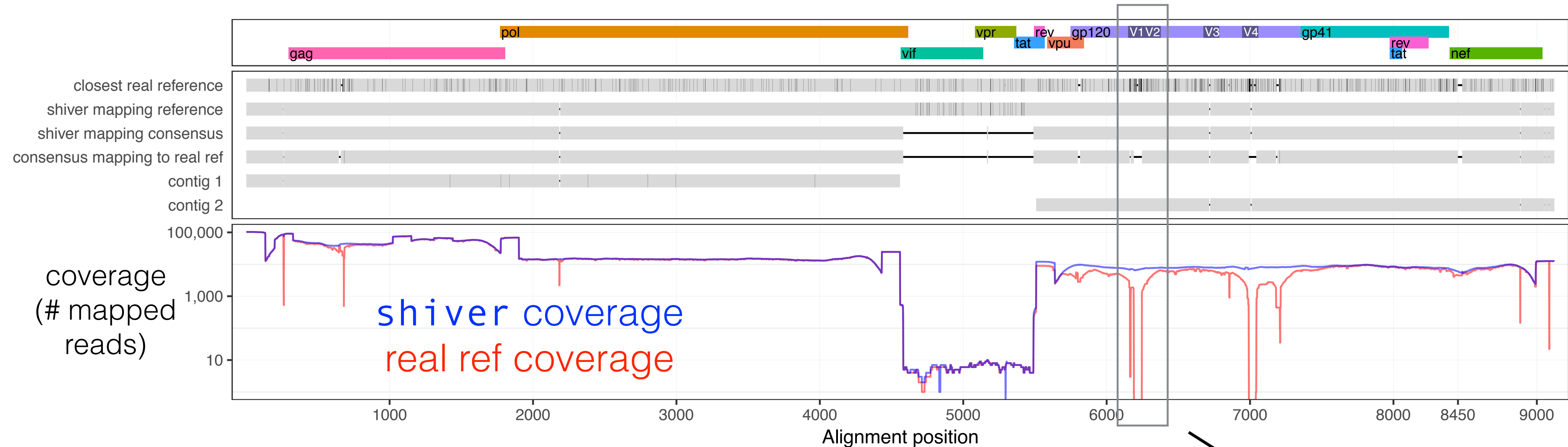
Also works on HCV, RSV, ...

```
$ shiver_align_contigs.sh
```

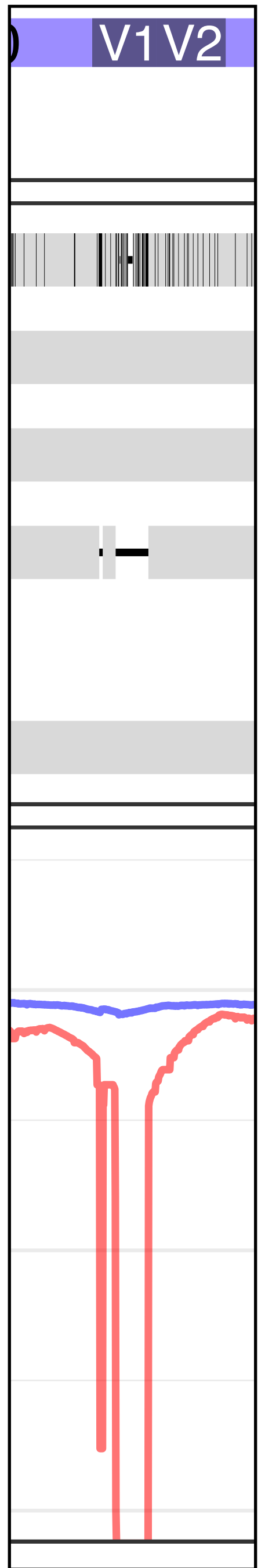
```
$ shiver_map_reads.sh
```



Map to *shiver*'s constructed reference,  
also map to the closest of 3259 real references (from database),  
compare.



Over 50 BEEHIVE samples + 65 public samples:  
median number of bases called differently with higher coverage:  
13; with lower coverage, 0.  
Recover missing sequence, often in *envelope* (vaccine design!).





# Application of `shiver` for 3 HIV projects:

Investigate the viral-genetic  
basis of virulence

BEEHIVE:  $N \sim 3,000$

PopART:  $N \sim 1,000$  currently, 9,000 eventually

PANGEA:  $N \sim 13,000$  currently, 23,000 eventually

Help interpret the  
effect of a population-  
level test-and-treat  
intervention in Zambia  
and South Africa

Generate new insights into  
transmission dynamics in  
generalized epidemics in Africa

Tanya  
Golubchik



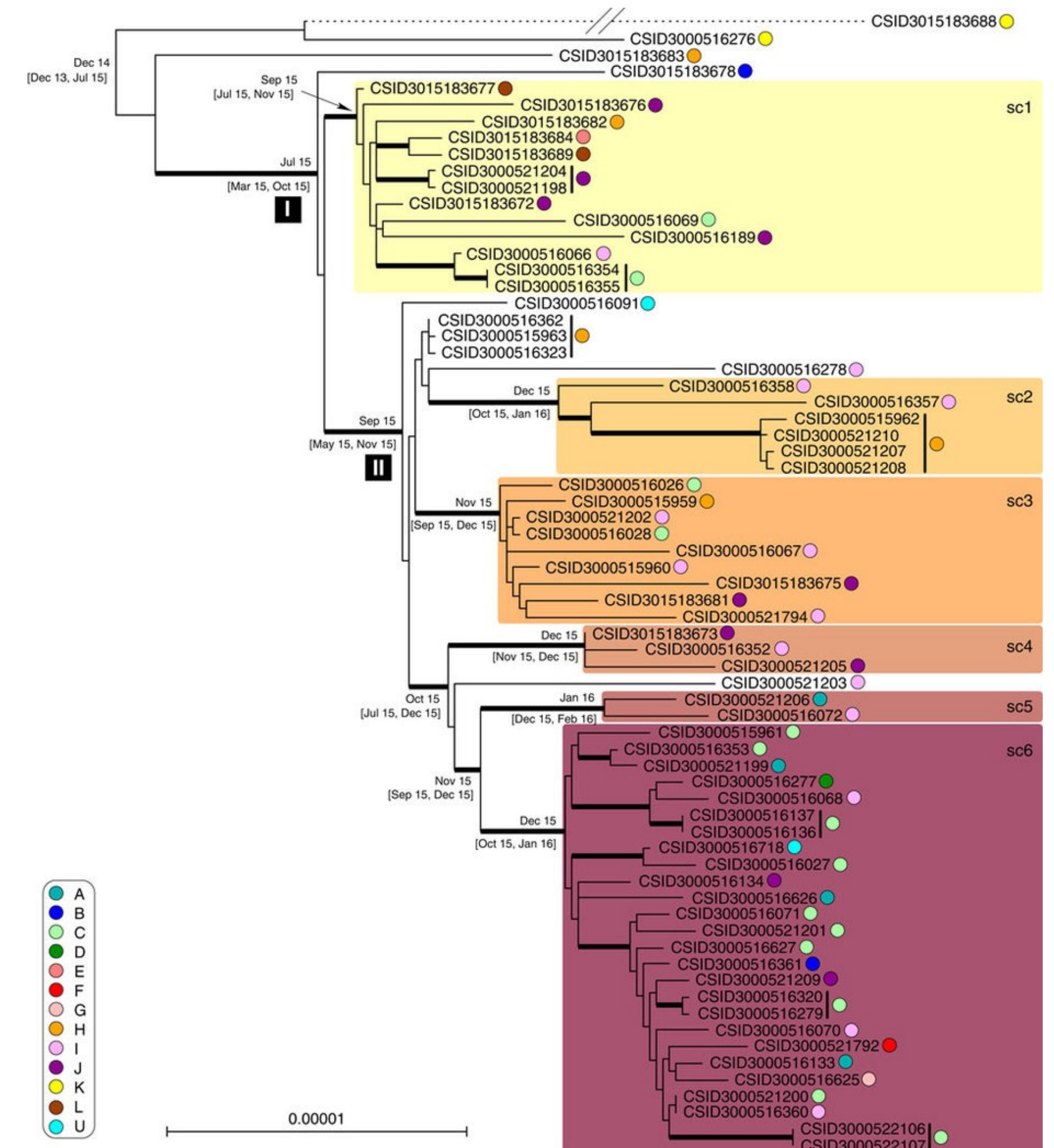
Problem: we want to identify risk factors for transmission. We need to identify not just transmission pairs or clusters, but *who infected whom*.

# Molecular epi with one pathogen sequence per individual

- Make a tree using one pathogen sequence per sampled individual.
- Define clusters: likely to be related by recent transmission.
- Investigate epidemiological correlates.

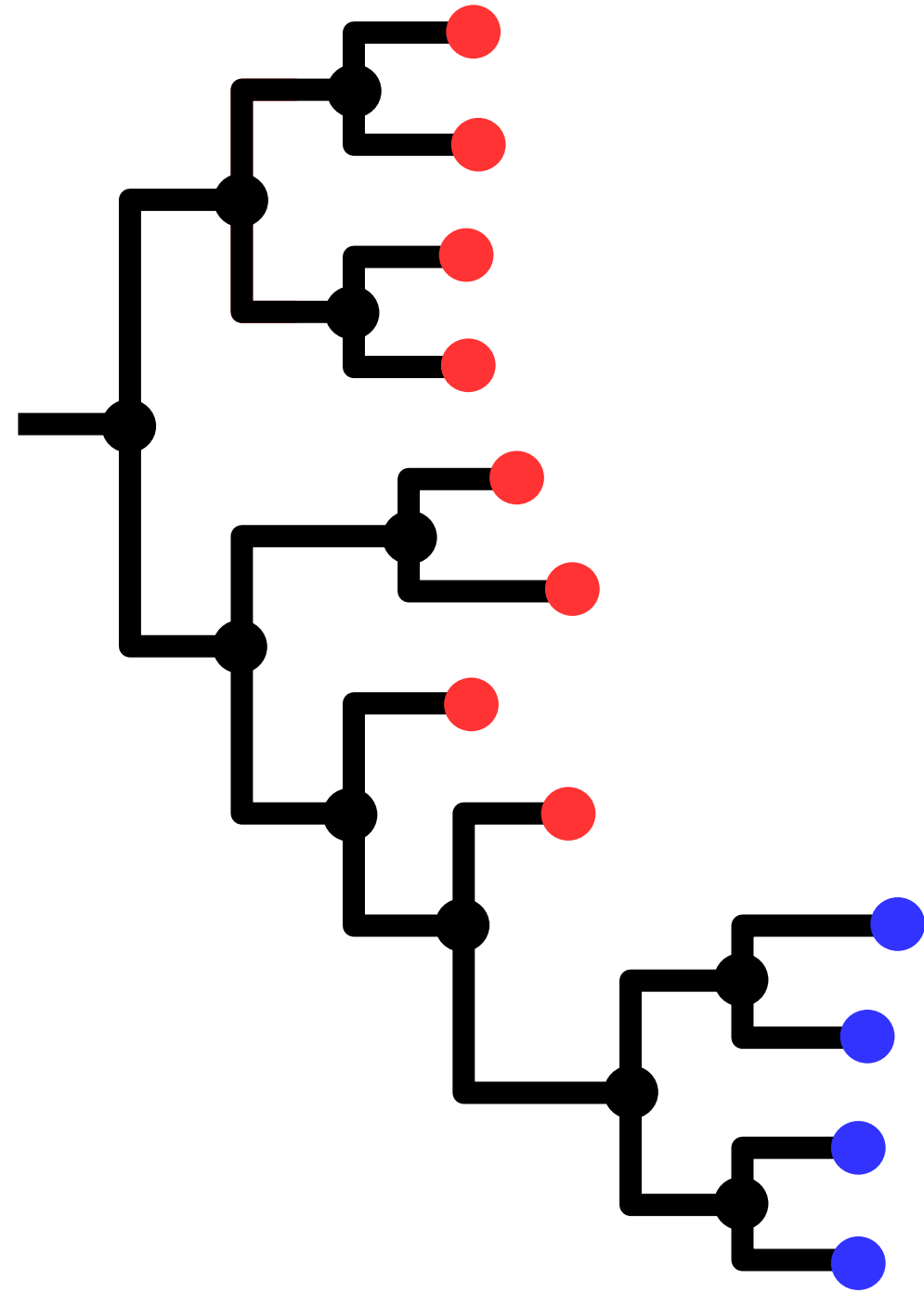
No information on who is transmitting!

Can do better by also using extra data + transmission model. Conclusions then dependent on both.

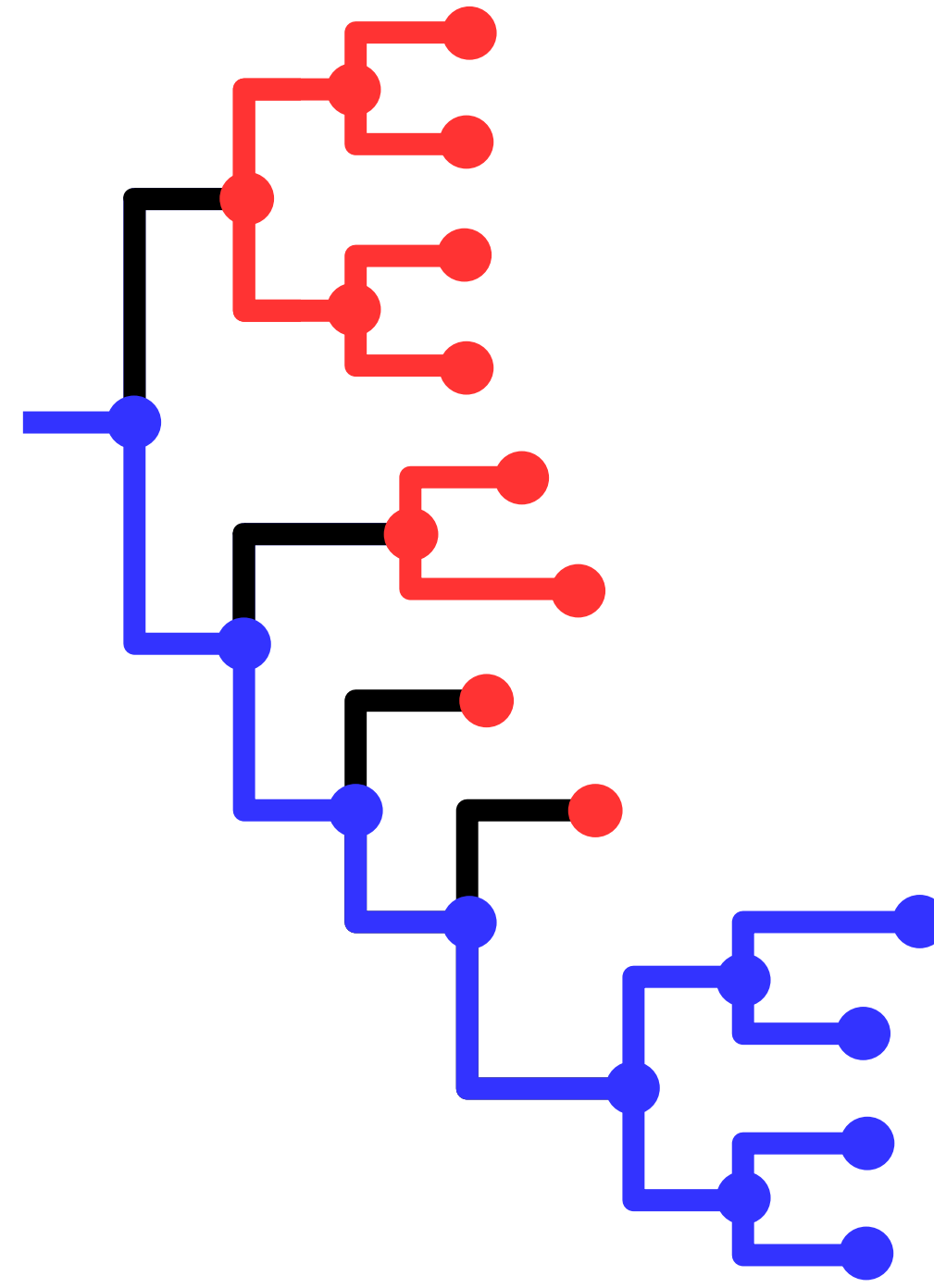




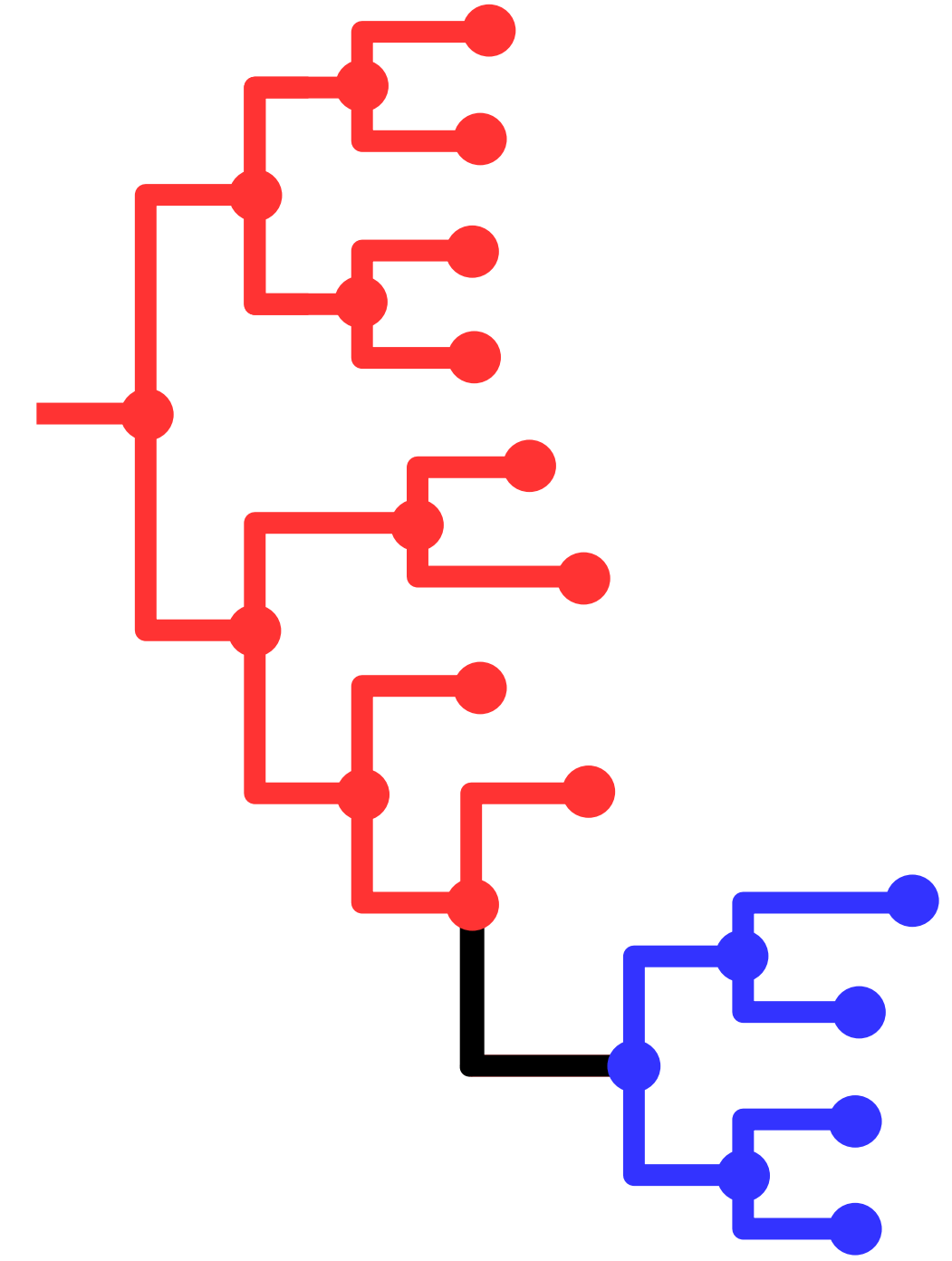
# Ancestral State Reconstruction



Known states (red or blue) at the tips.



Red evolved from blue: 4+ changes of state needed.

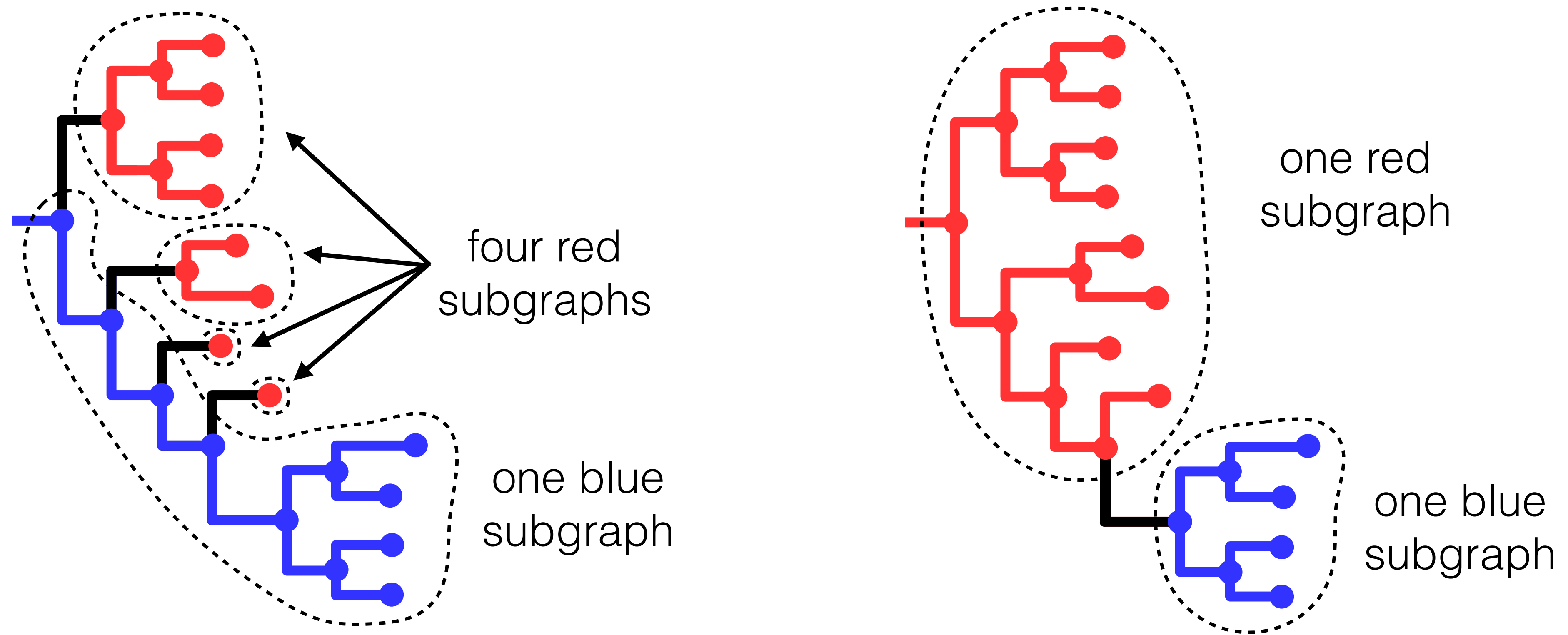


Blue evolved from red: 1 change of state needed.

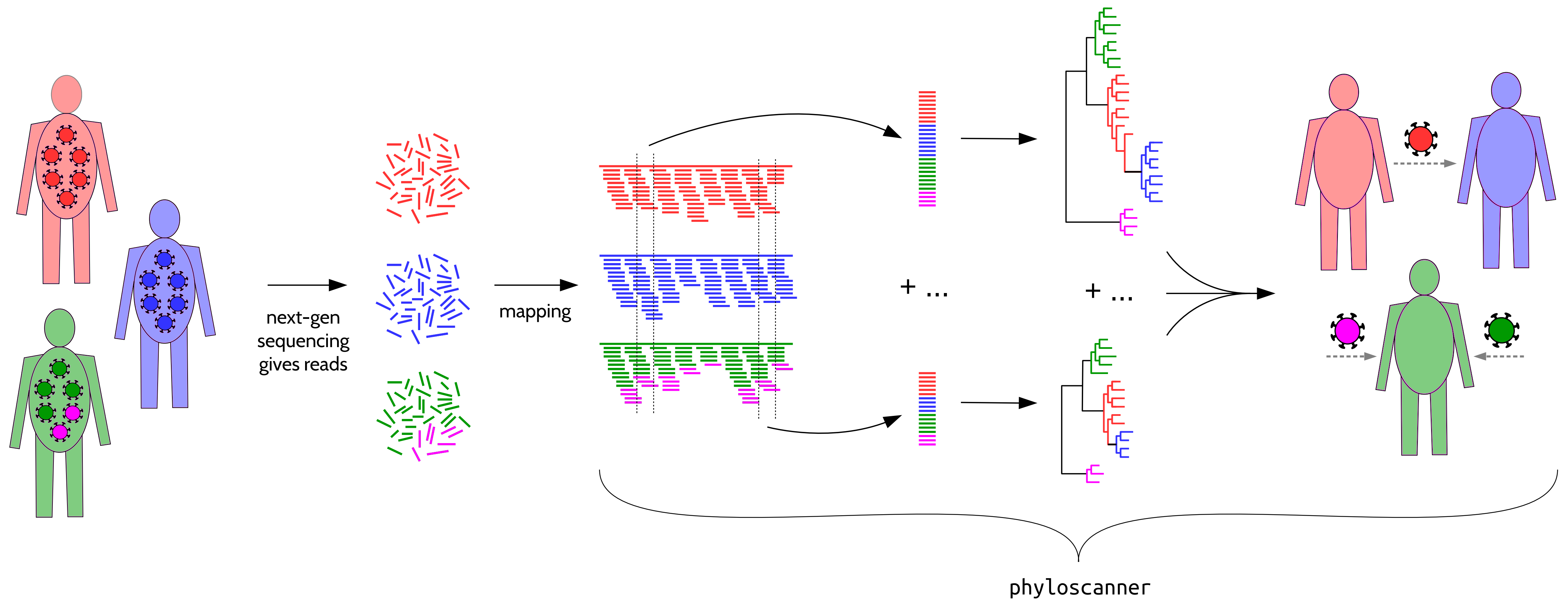
e.g. colour for “which person was this virus in?” - Romero-Severson et al. PNAS 2016



# Ancestral State Reconstruction: subgraphs



*A subgraph* is connected/continuous region of the phylogeny. Later, we'll define these regions by the state assigned to them, i.e. one subgraph = one solid block of colour.

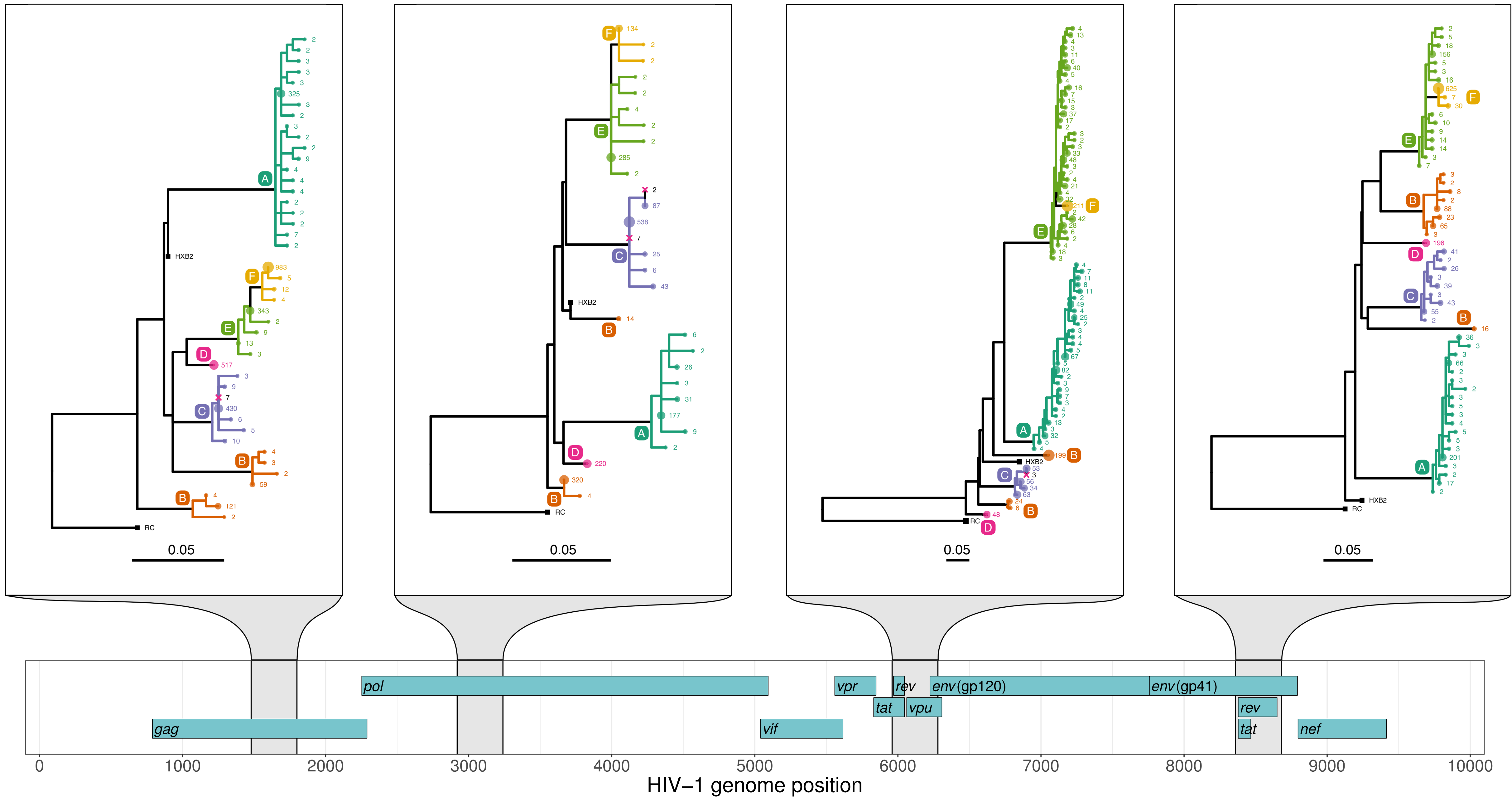


Wymant, Hall *et al.*, MBE 2017

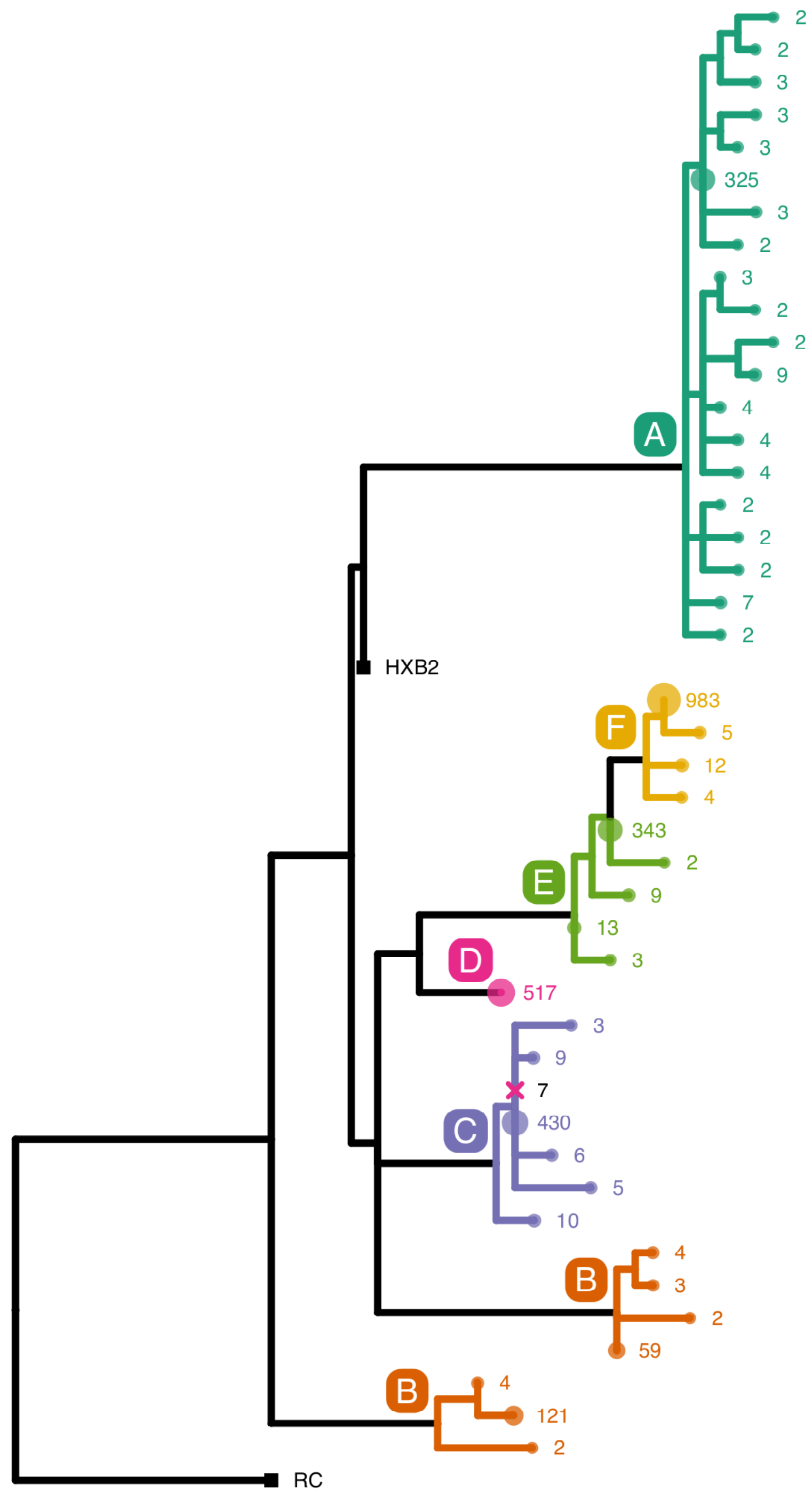
Alignment with MAFFT: Katoh *et al.*, Nucleic Acids Res. 2002

Phylogenetic inference with RAxML: Stamatakis, Bioinformatics 2014



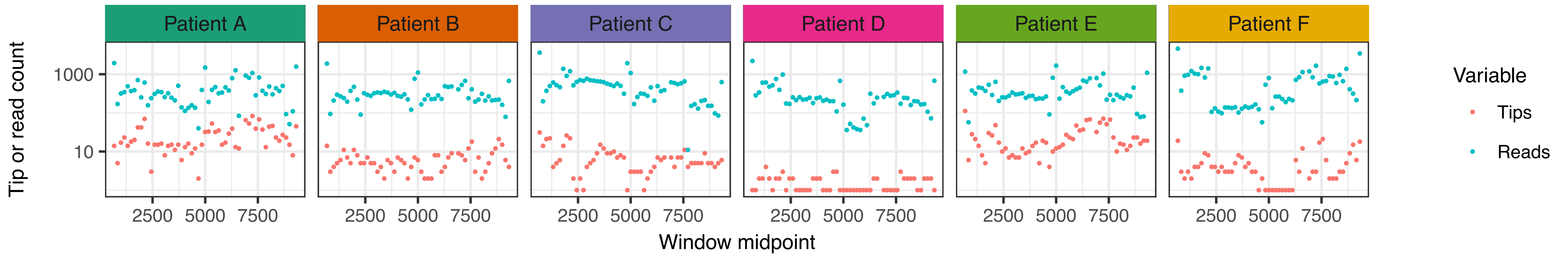








# Per-patient summary statistics along the genome

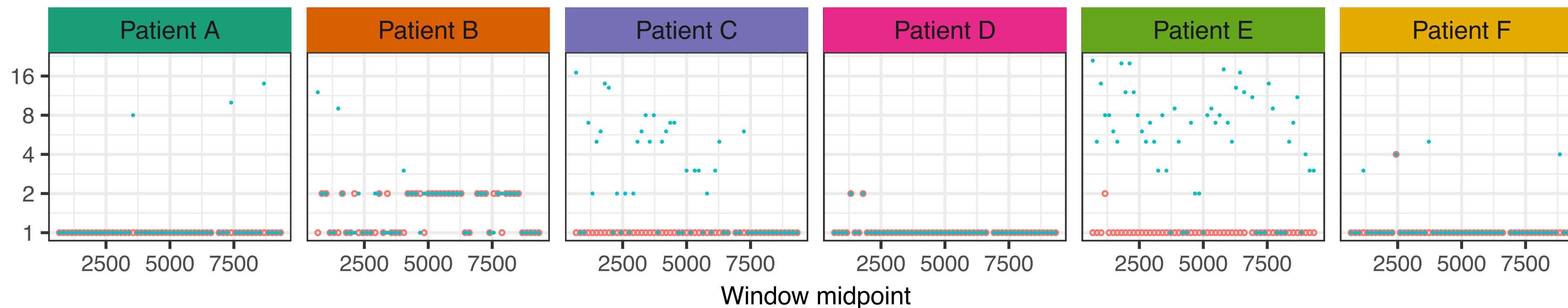


100-1000 reads,  
10-100 unique reads.  
Diverse!



100-1000 reads,  
1-2 unique reads.  
Not diverse!

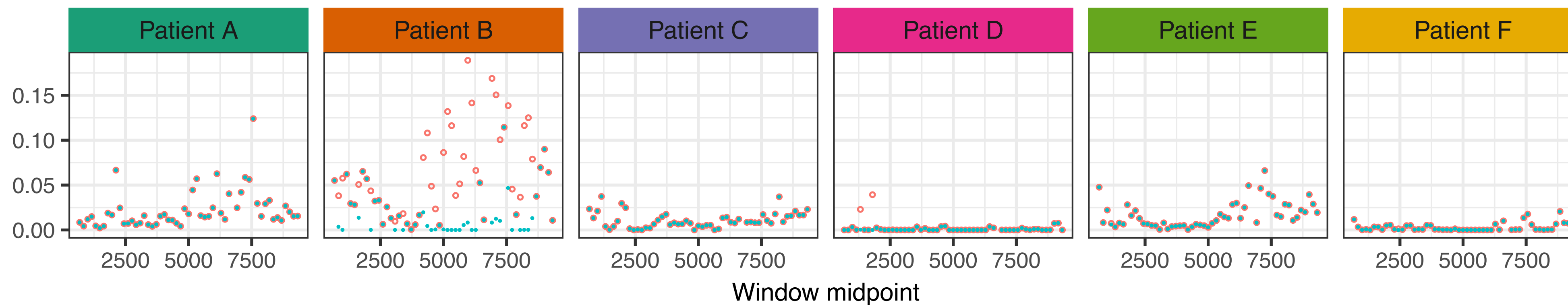
Subgraph or clade count



Variable

- Subgraphs
- Clades

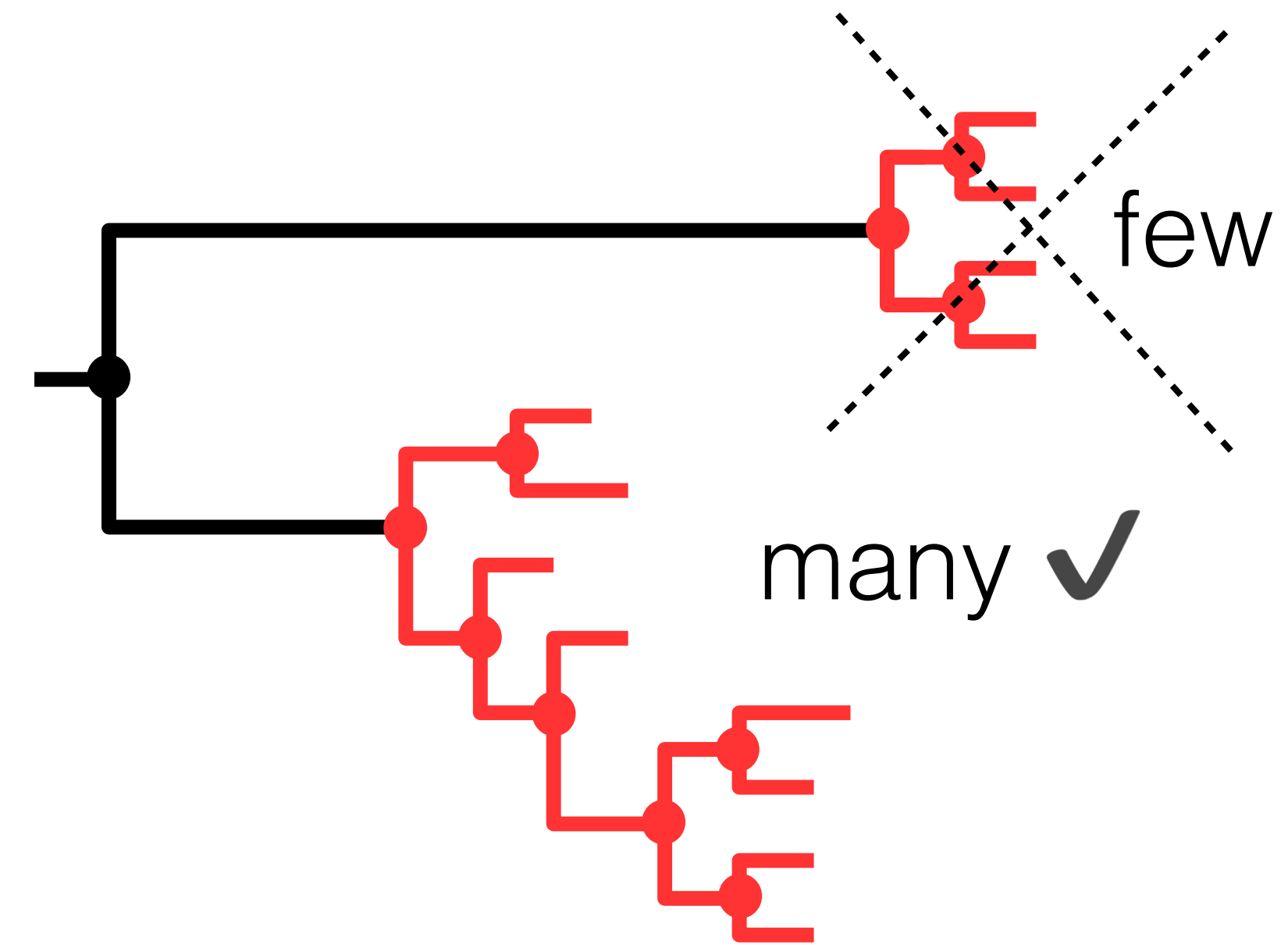
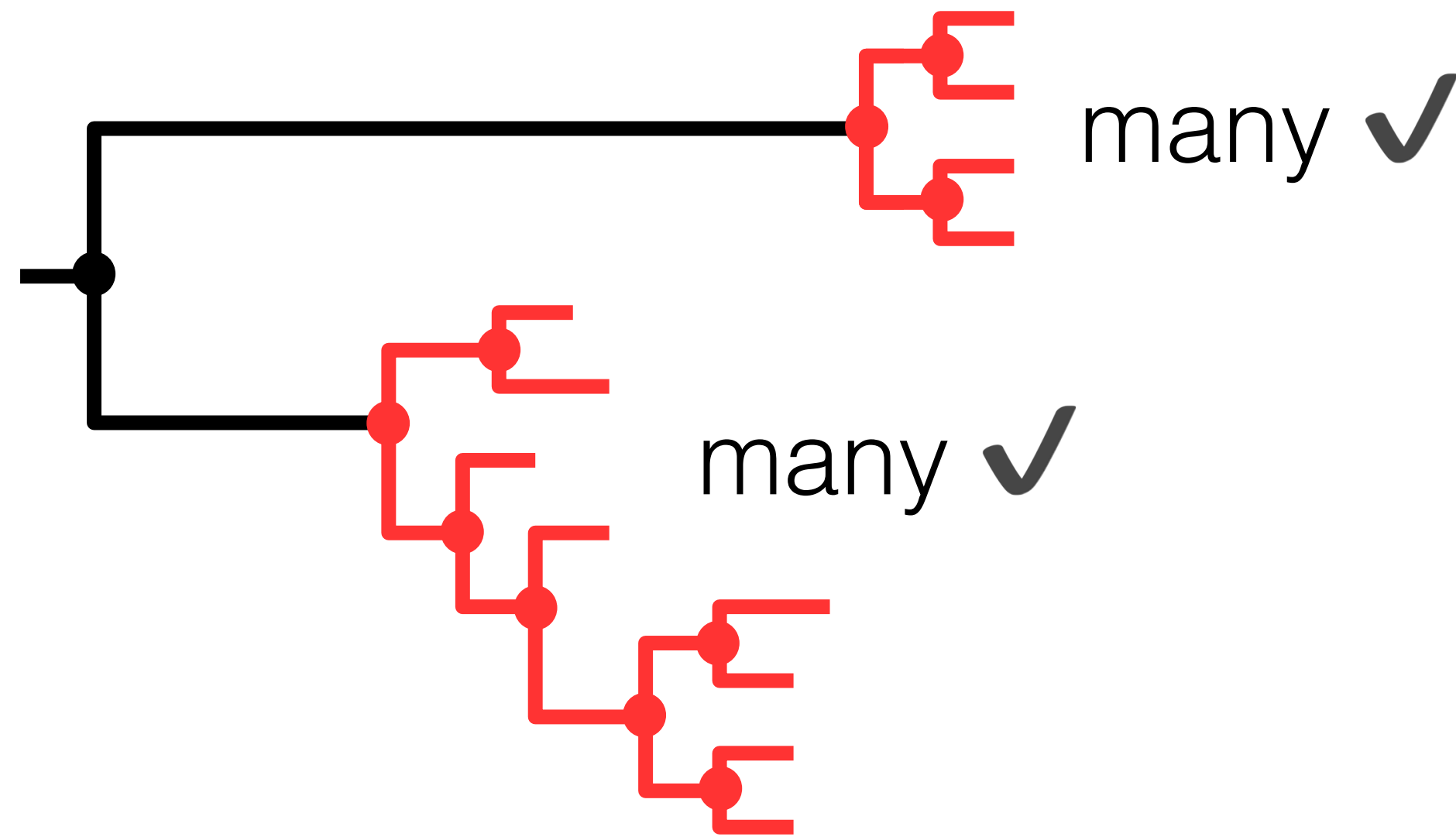
Mean root-to-tip-distance  
(read-weighted)



Tip set

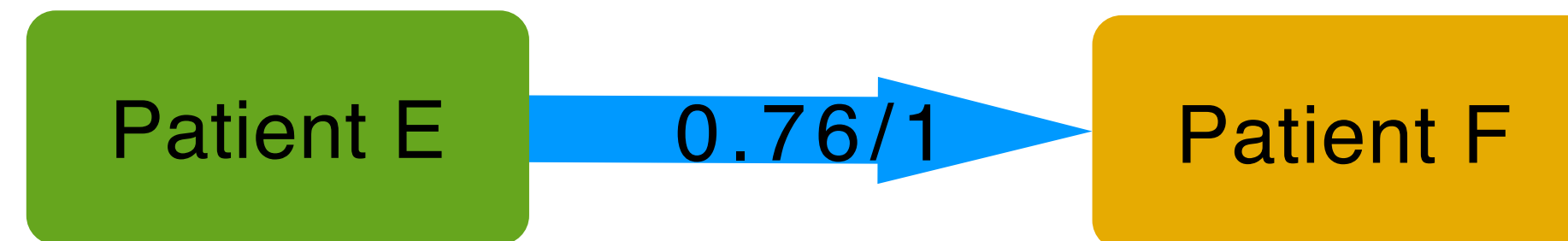
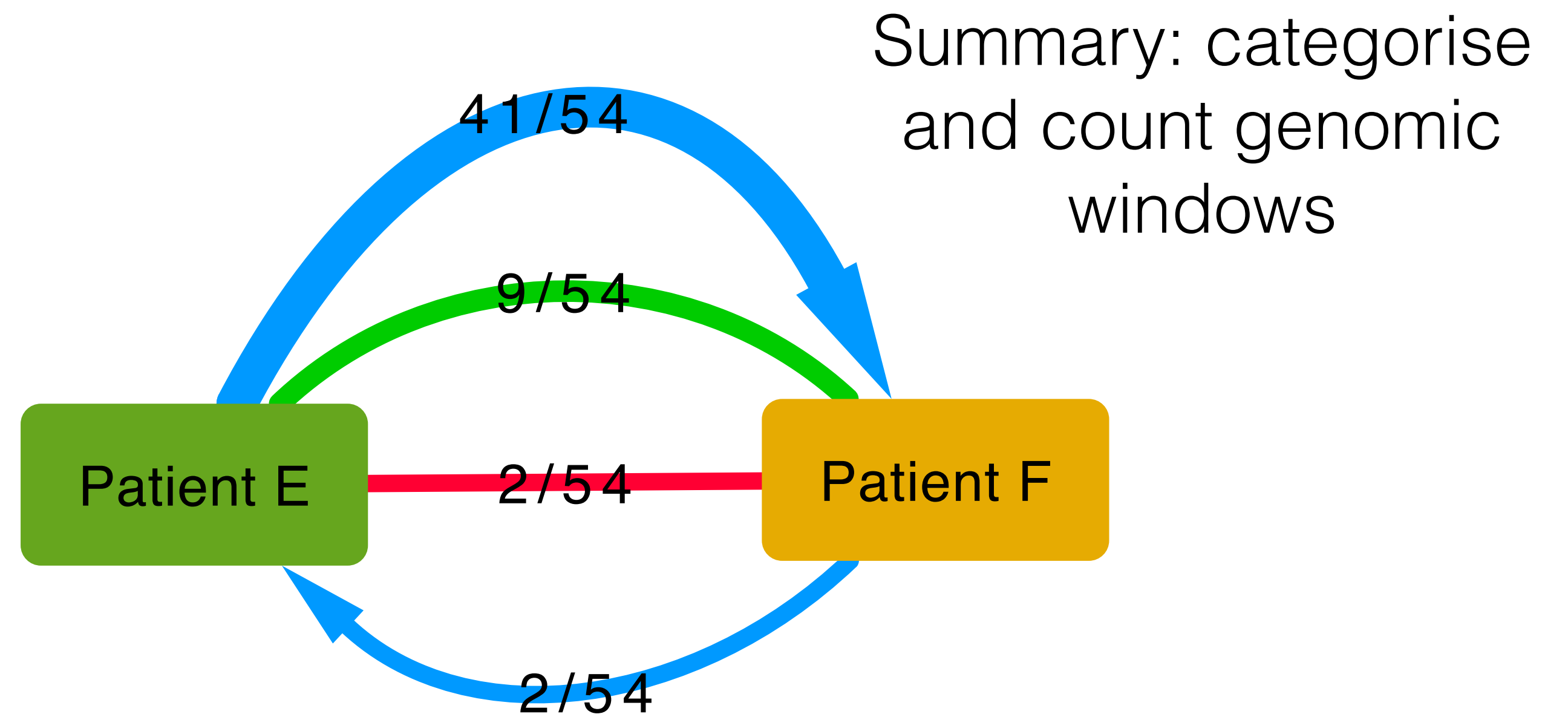
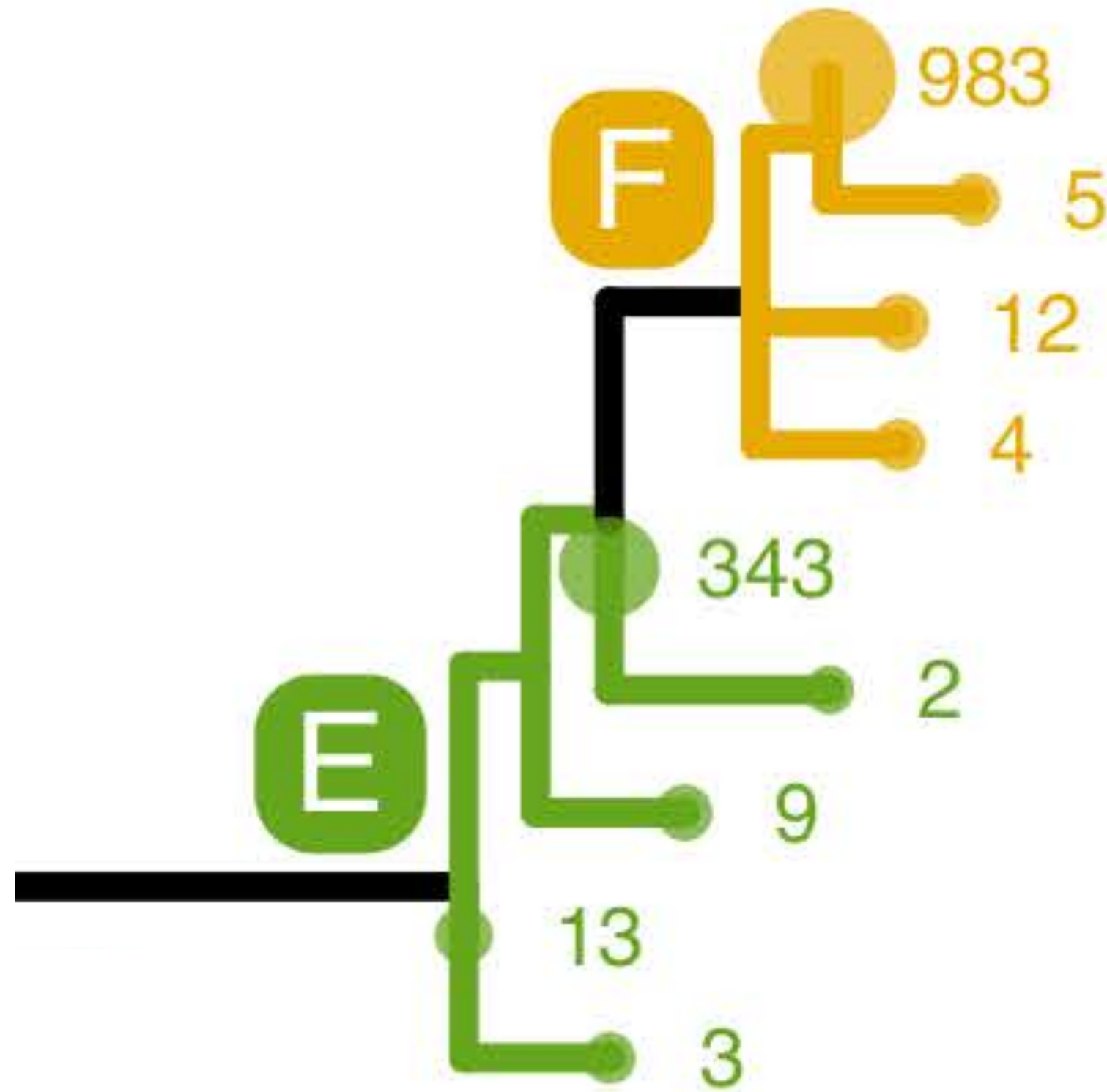
- All
- Largest subgraph

Dually infected individual or contaminated sample?



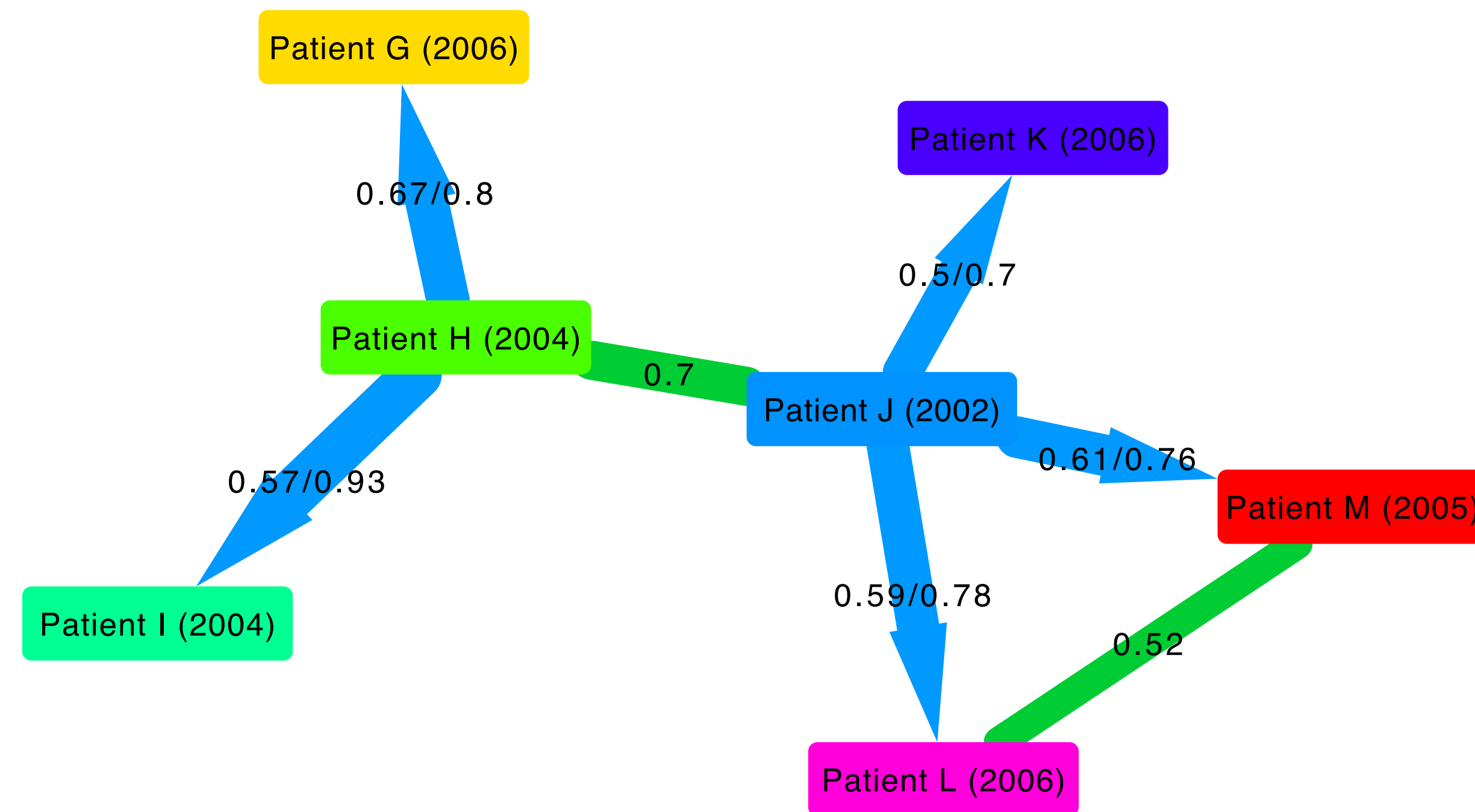
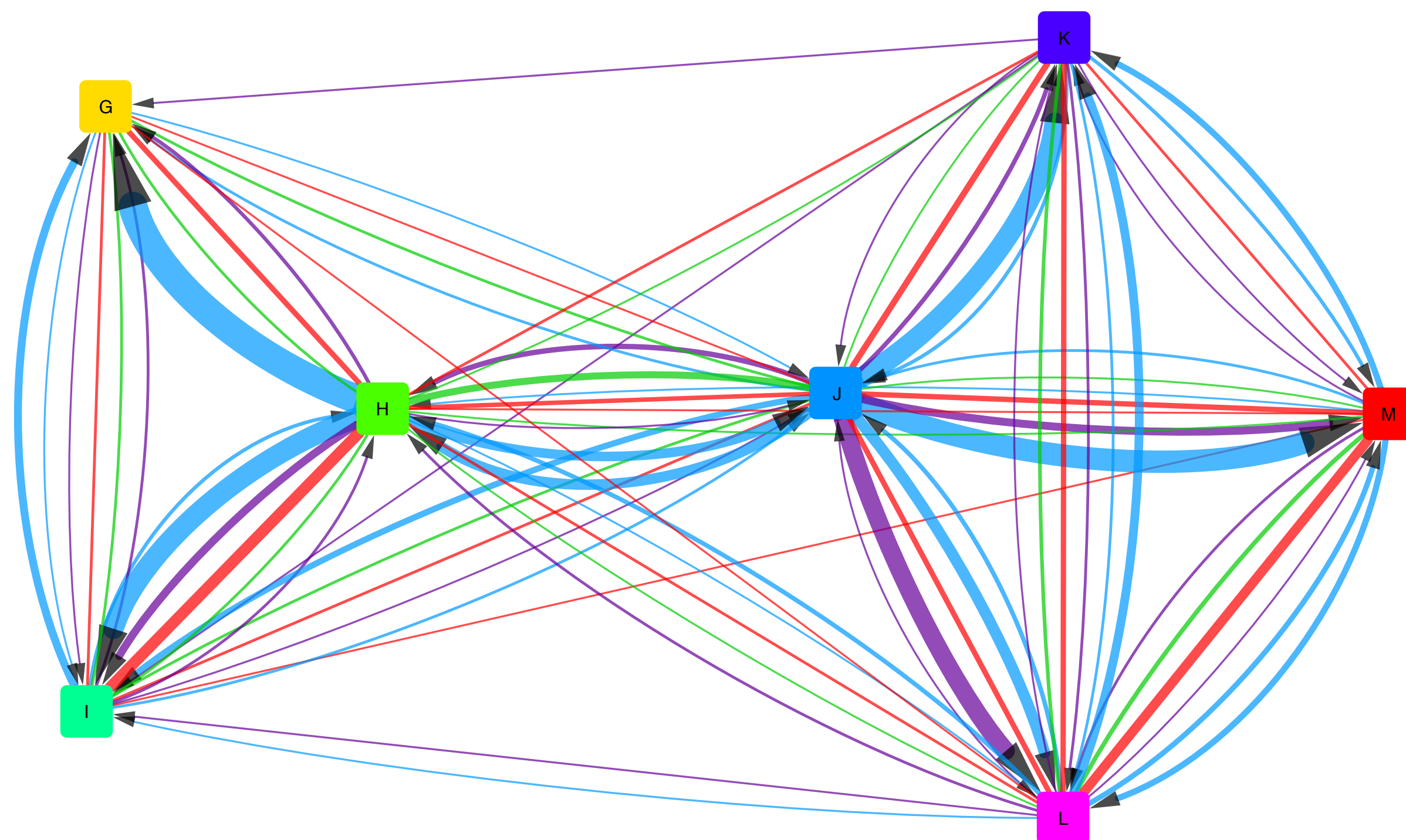


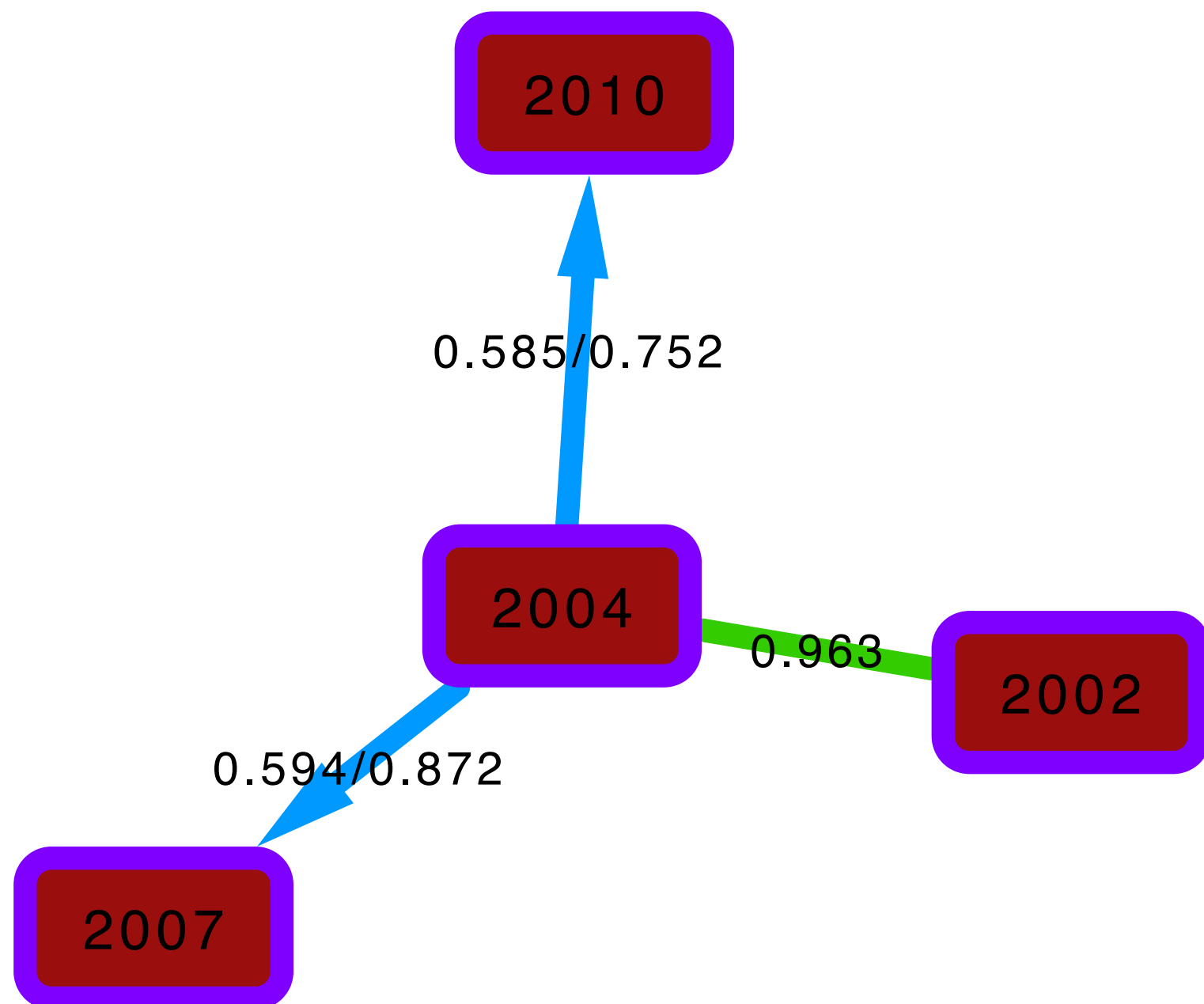
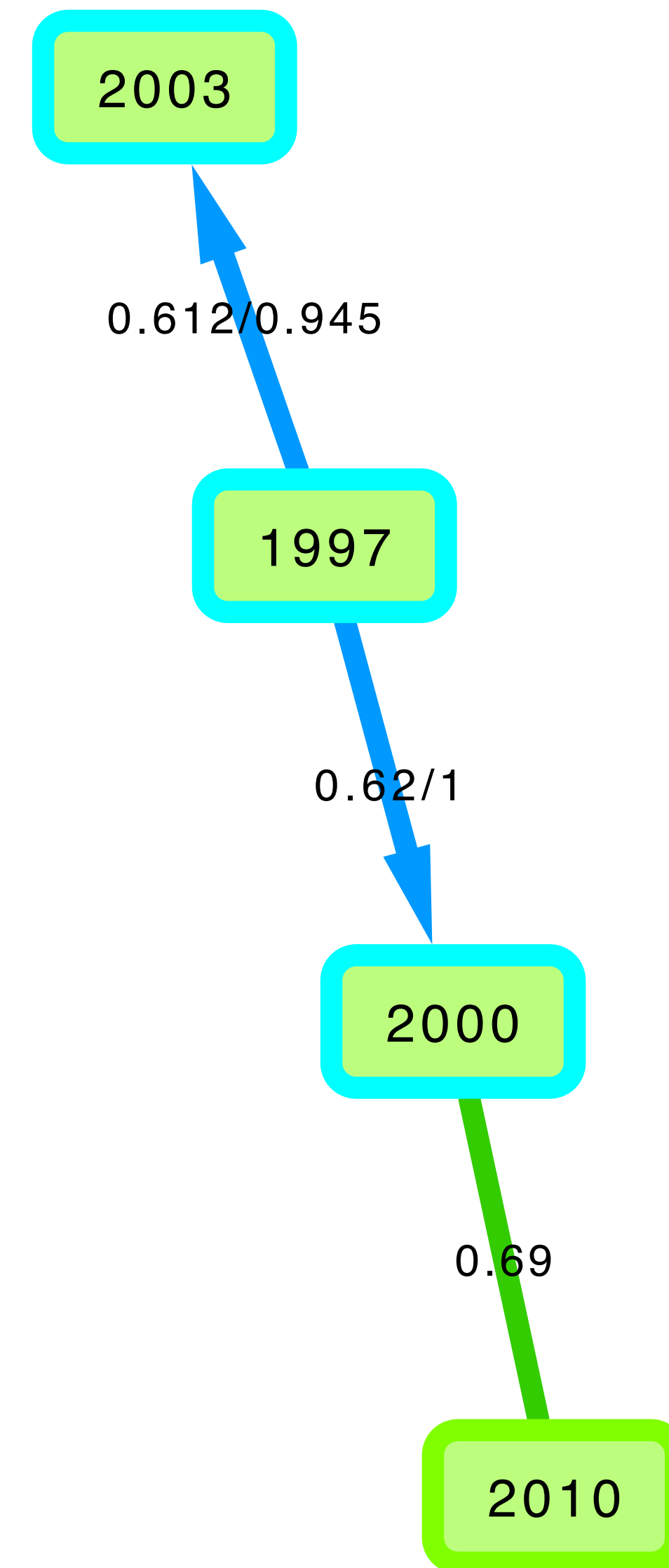
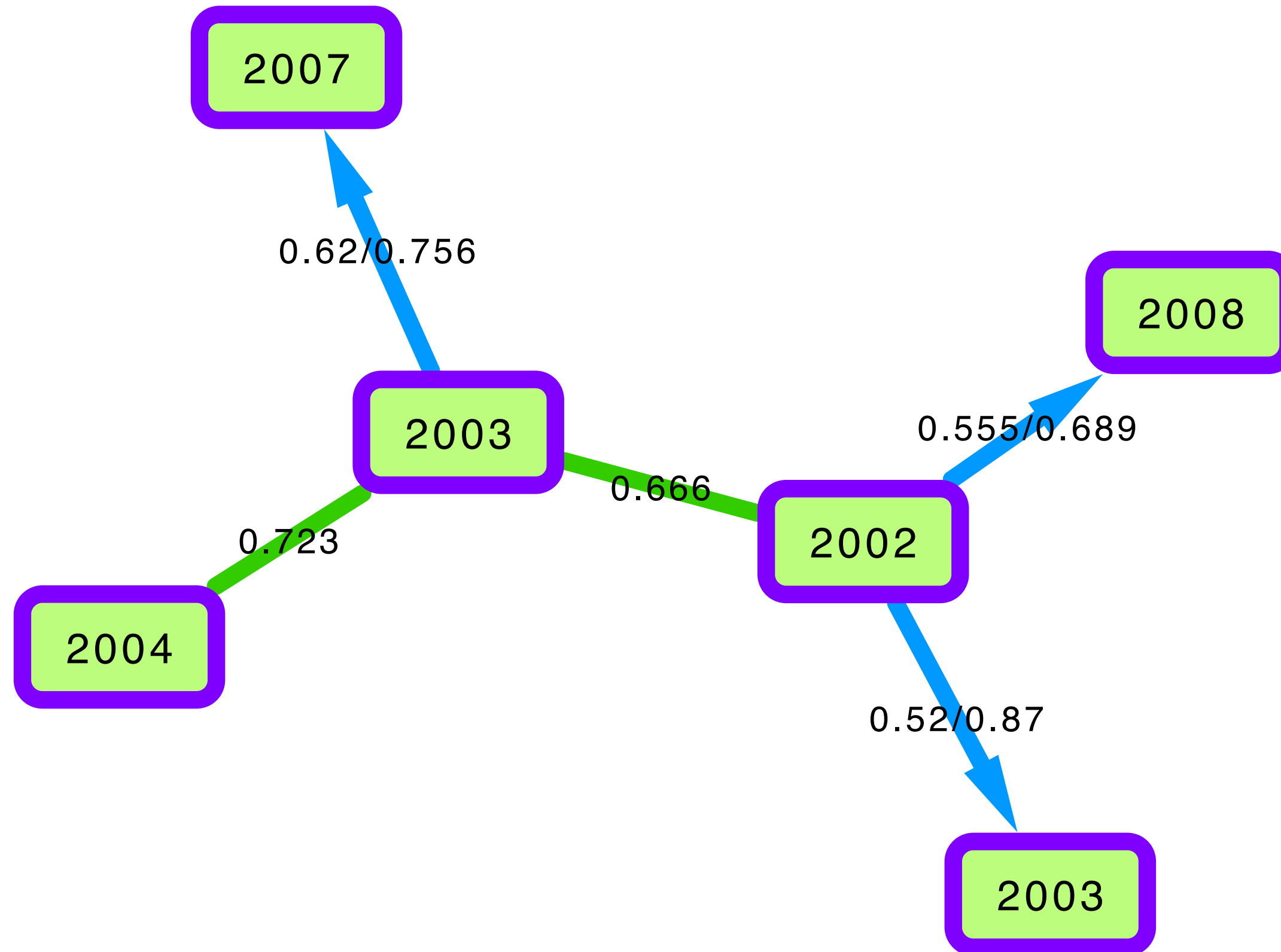
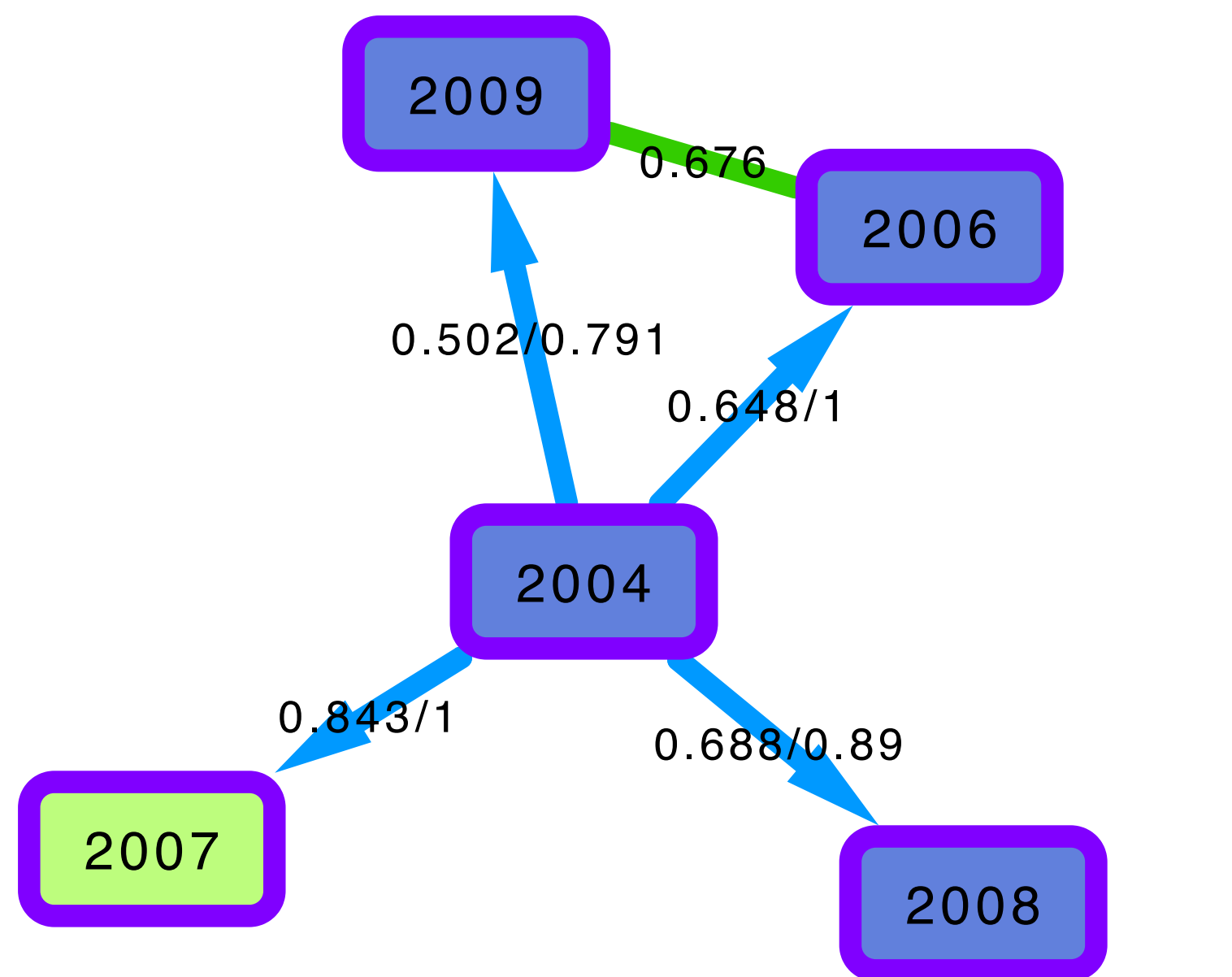
# Transmission



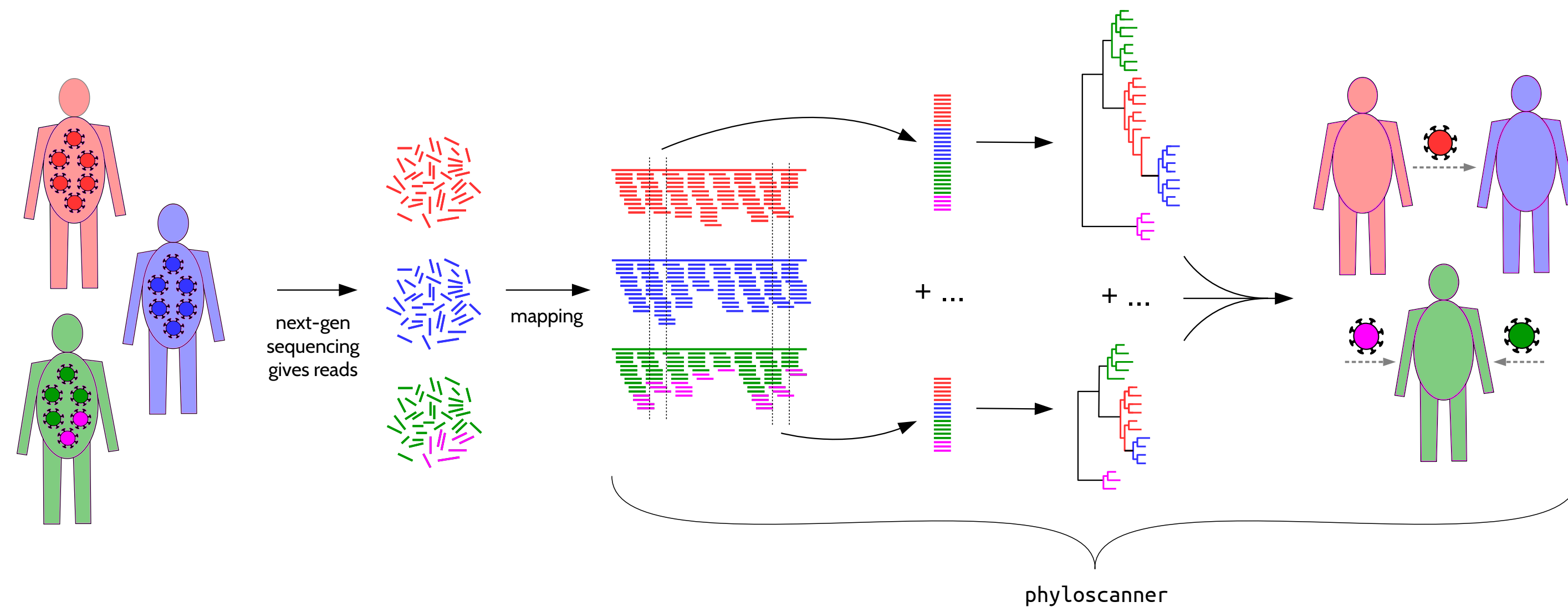
Simpler summary: aggregate categories







# Summary



phyloscanner creates within- & between-host phylogenies along the genome from NGS reads. Analysing these, or other such phylogenies, it identifies

- **transmission**: one individual's pathogen population being ancestral to someone else's
- **multiple infection**: distinct subpopulations in the same host
- **contamination**: exact duplicates, phylogenetic outliers
- **recombination**

Wymant, Hall *et al.* MBE 2017

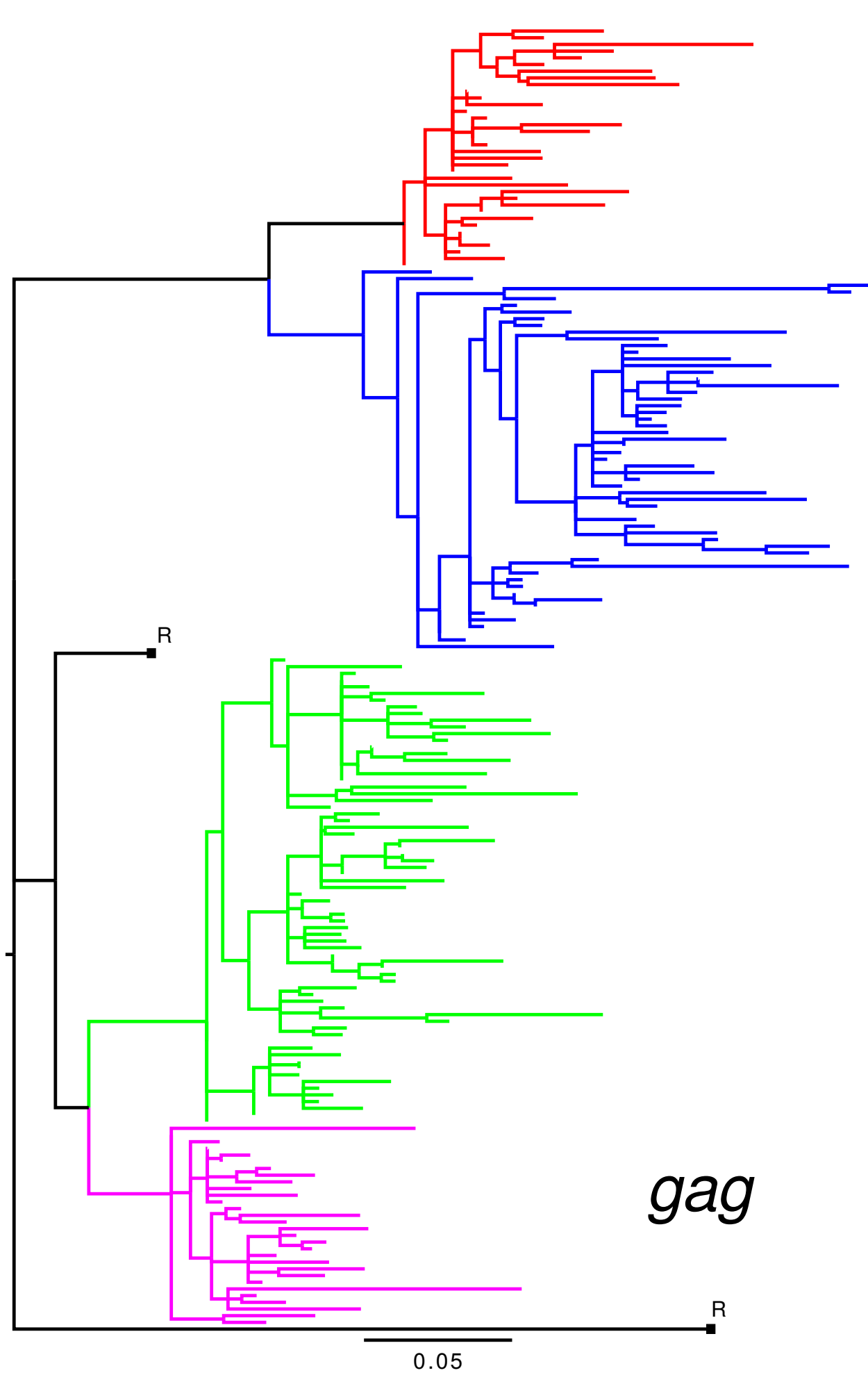
[GitHub.com/BDI-pathogens/phyloscanner](https://github.com/BDI-pathogens/phyloscanner)



Extra slides: application to other  
sequencing platforms and pathogens,  
and using phyloscanner



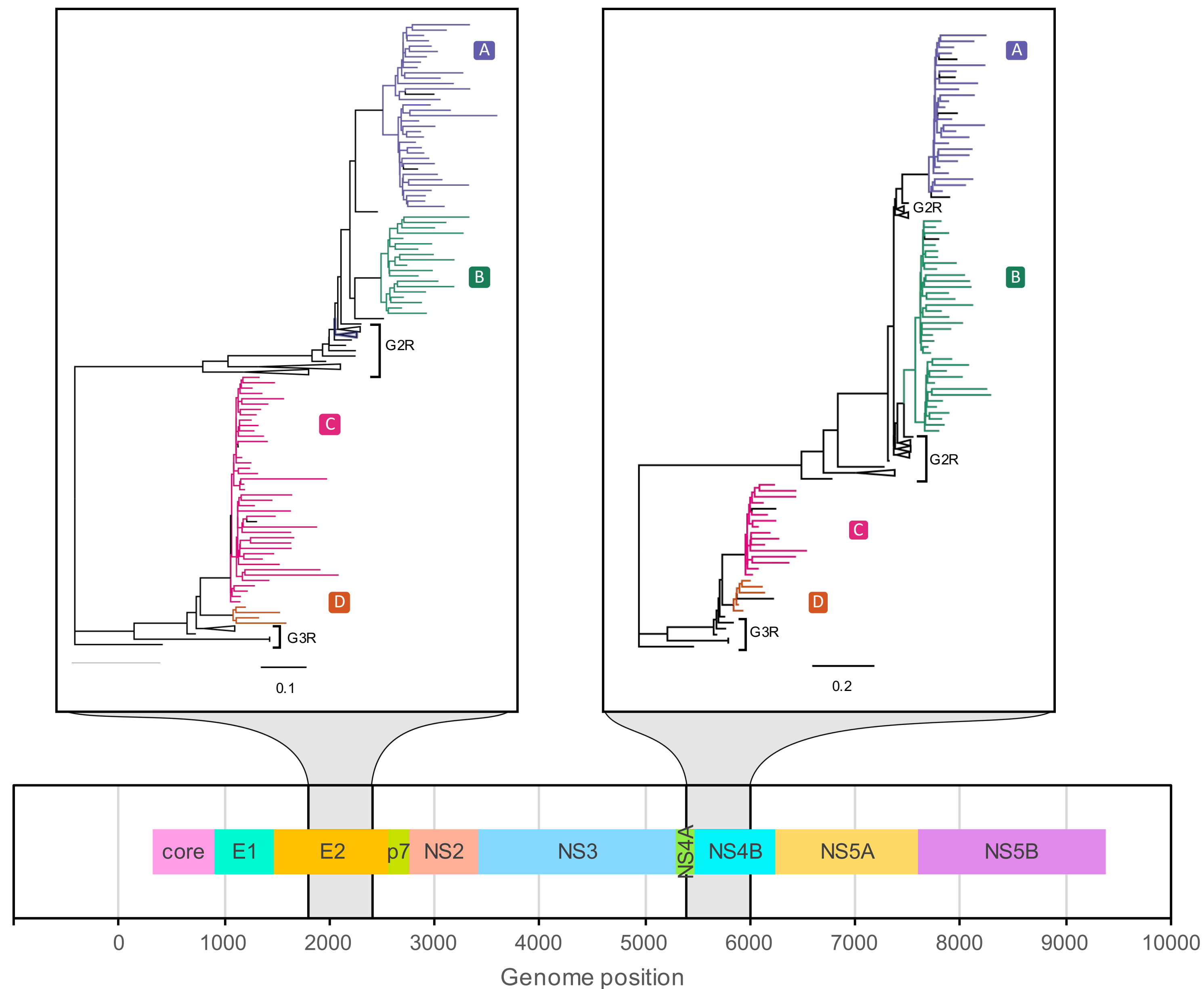
# Four more BEEHIVE patients sequenced with Roche 454



Tanya Golubchik



# Hepatitis C Virus Sequenced With Oxford Nanopore



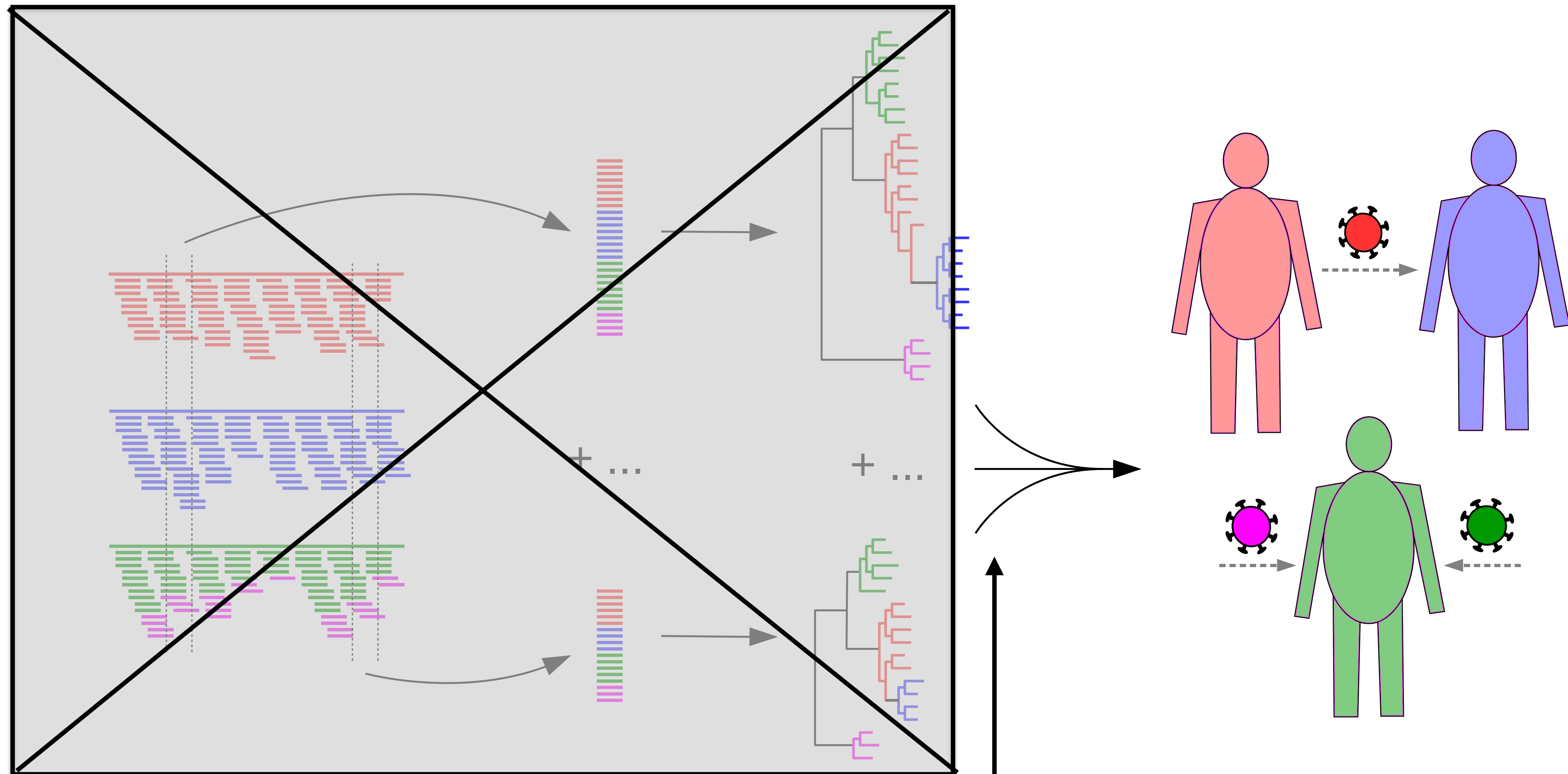
- The Stop-HCV Consortium:**
- |                 |                    |
|-----------------|--------------------|
| Eleanor Barnes, | Diana Koletzki     |
| Jonathan Ball   | Natasha Martin     |
| Diana Brainard  | Benedetta Massetto |
| Gary Burgess    | Tamyo Mbisa        |
| Graham Cooke    | John McHutchison   |
| John Dillon     | Jane McKeating     |
| Graham R Foster | John McLauchlan    |
| Charles Gore    | Alec Miners        |
| Neil Guha       | Andrea Murray      |
| Rachel Halford  | Peter Shaw         |
| Cham Herath     | Peter Simmonds     |
| Chris Holmes    | Chris C A Spencer  |
| Anita Howe      | Paul Targett-Adams |
| Emma Hudson     | Emma Thomson       |
| William Irving  | Peter Vickerman    |
| Salim Khakoo    | Nicole Zitzmann    |
| Paul Klenerman  |                    |



David Bonsall



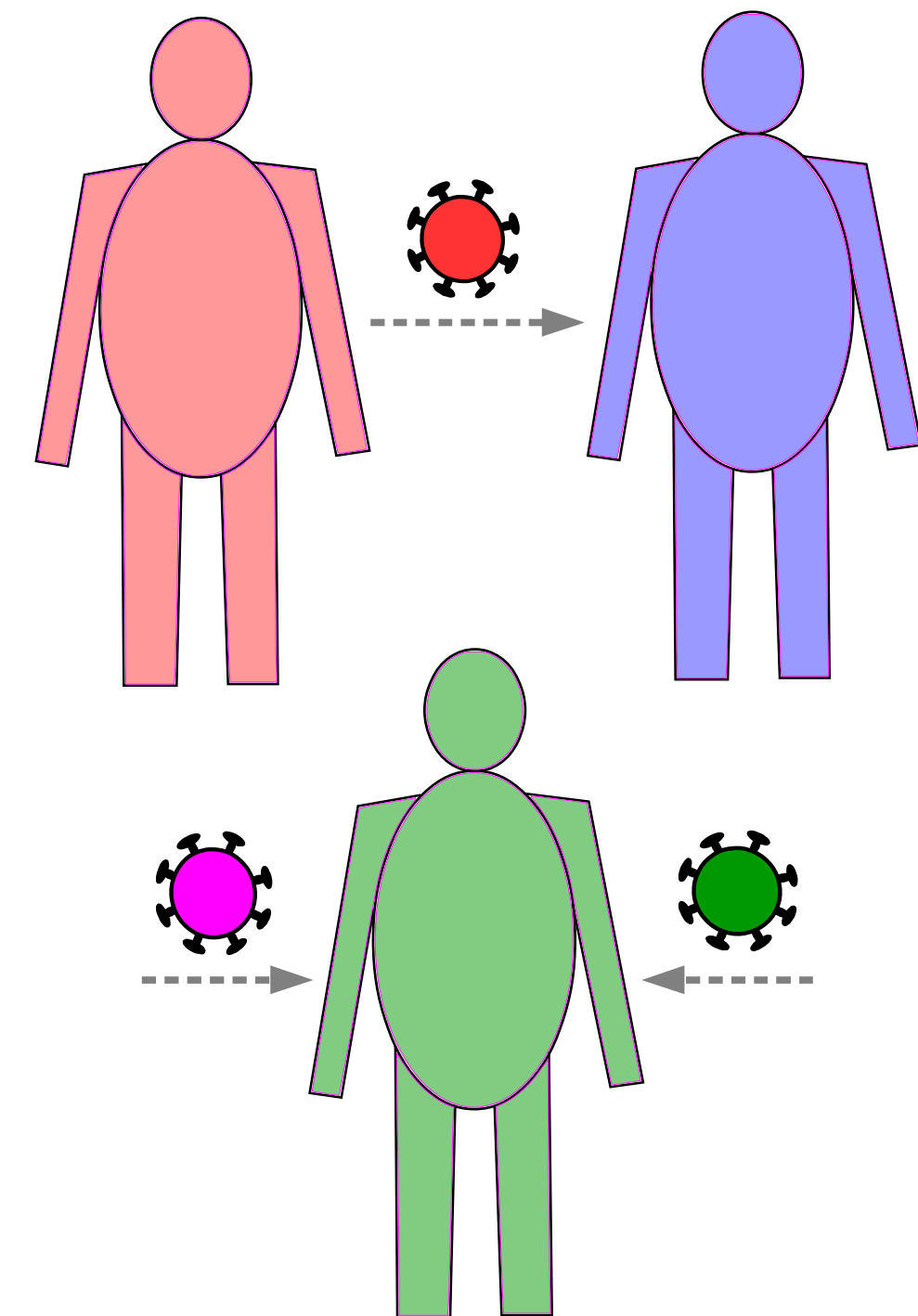
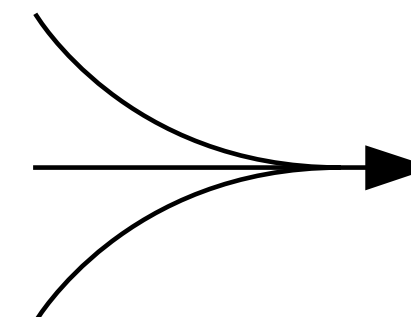
Mariateresa  
de Cesare



Sequence data from Croucher  
*et al.* PLoS Bio. 2016: multiple  
 colony picks per carrier of  
*S. pneumoniae*



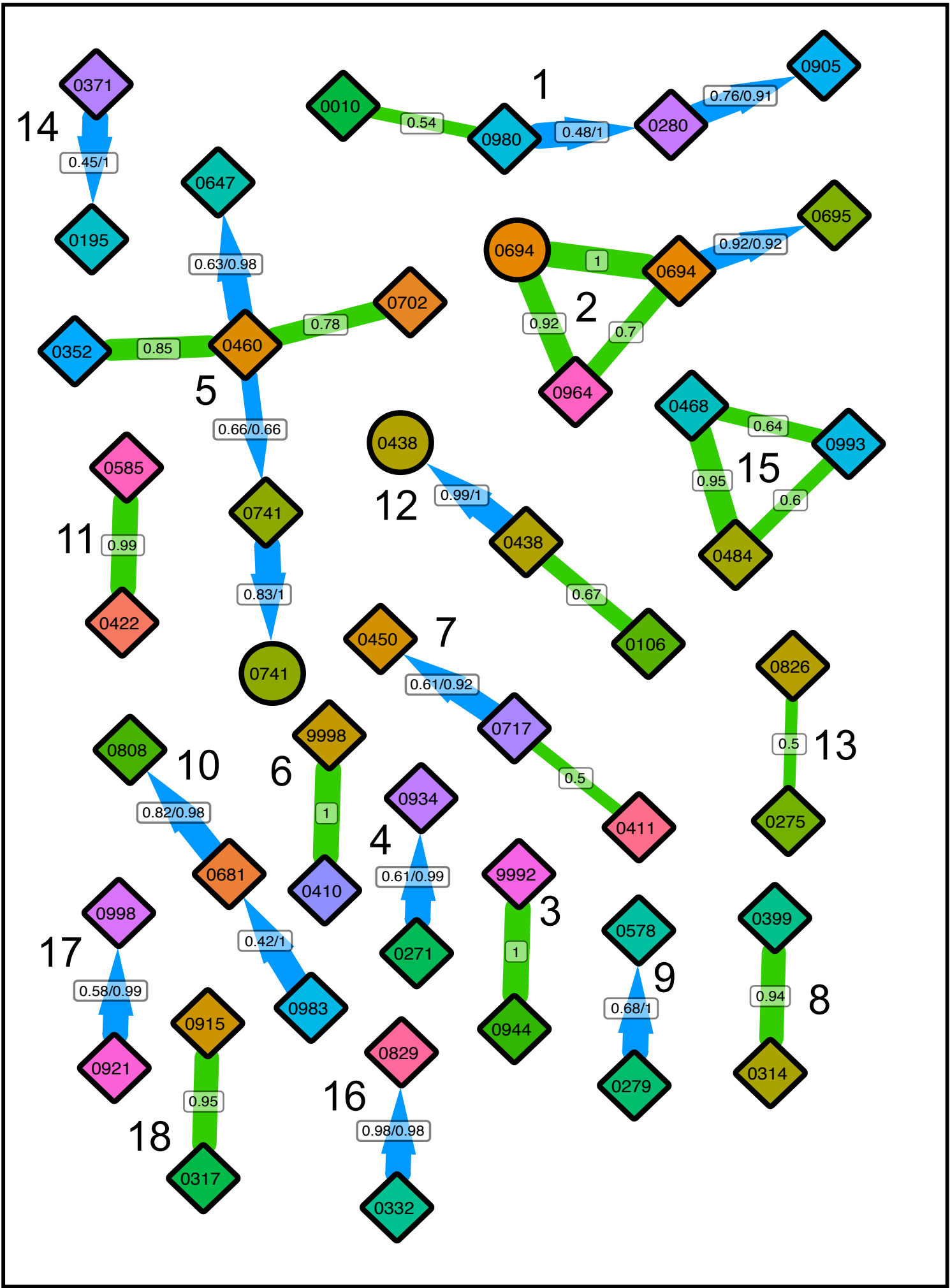
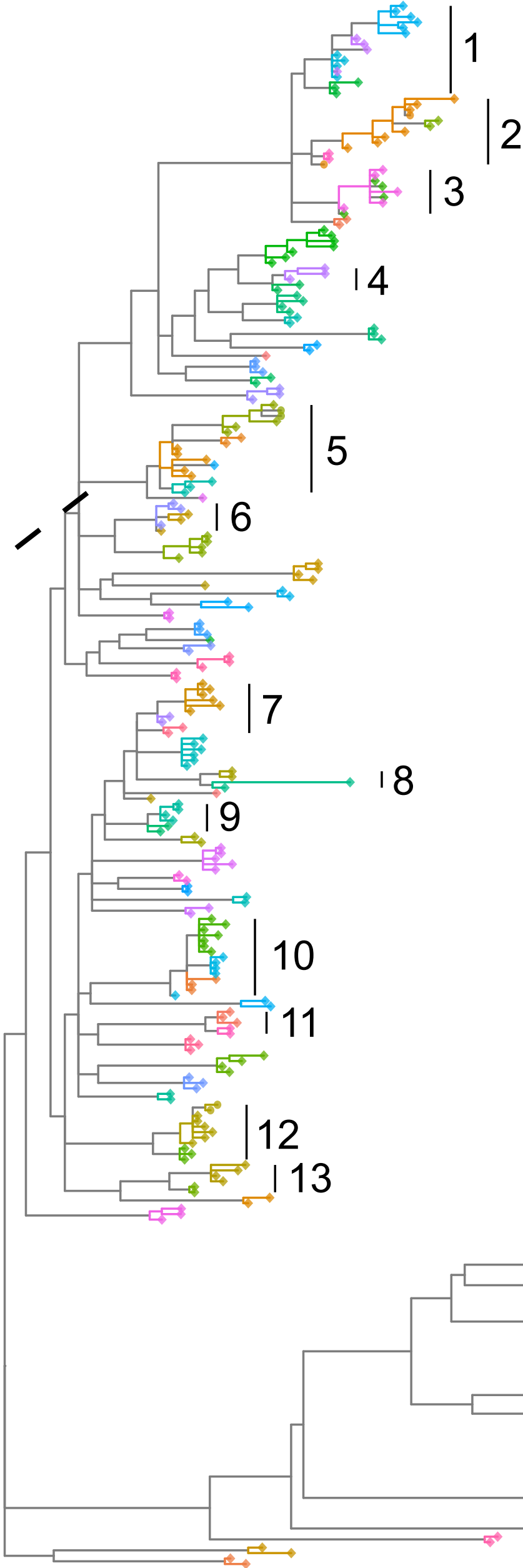
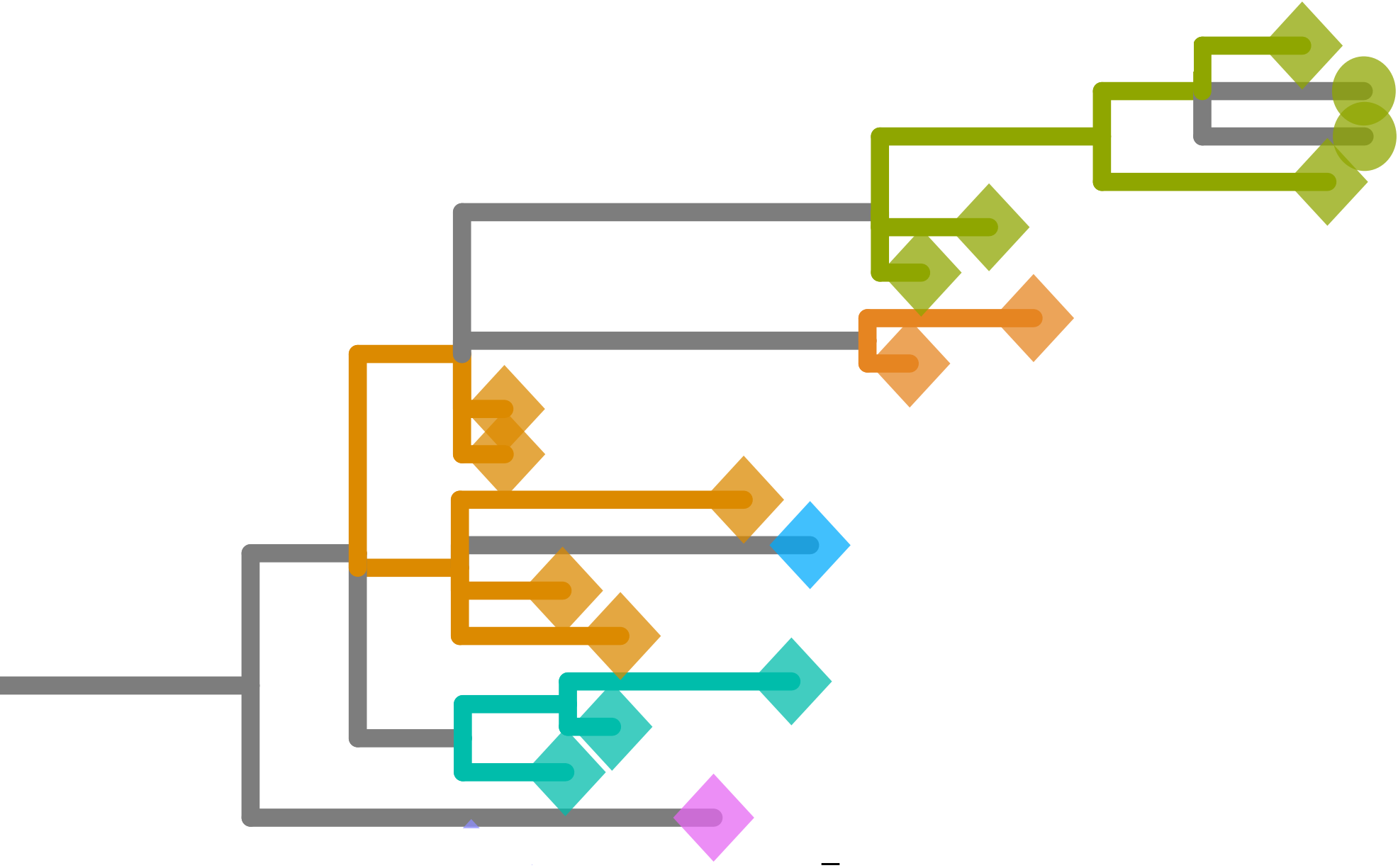
100 posterior trees  
 with MrBayes



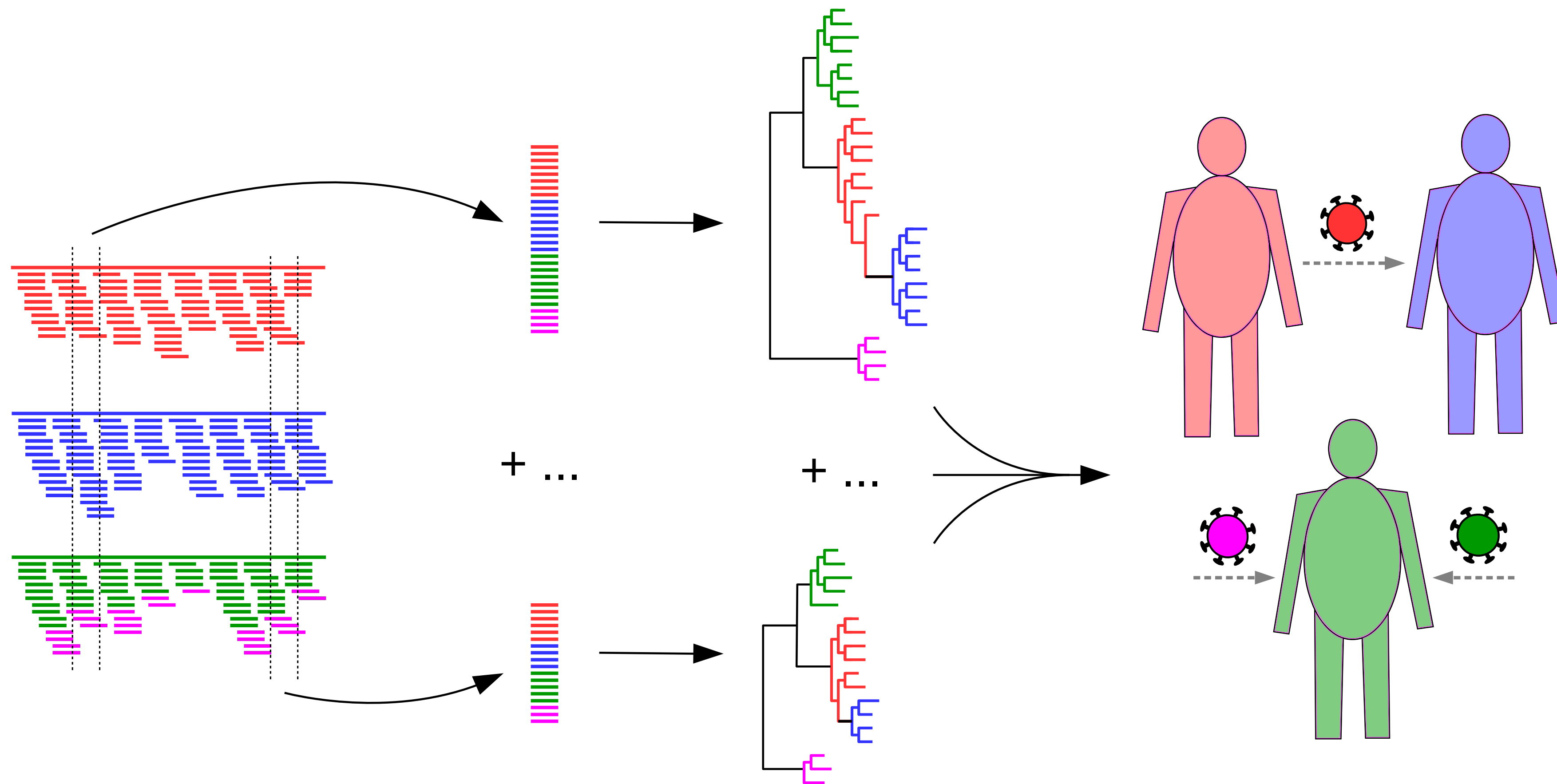


**The Maela Pneumococcal  
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Nicholas J. Croucher  
Simon Harris  
Jukka Corander  
David Goldblatt  
Julian Parkhill  
Francois Nosten  
Claudia Turner  
Paul Turner



Using phyloscanner: [GitHub.com/BDI-pathogens/phyloscanner](https://github.com/BDI-pathogens/phyloscanner)

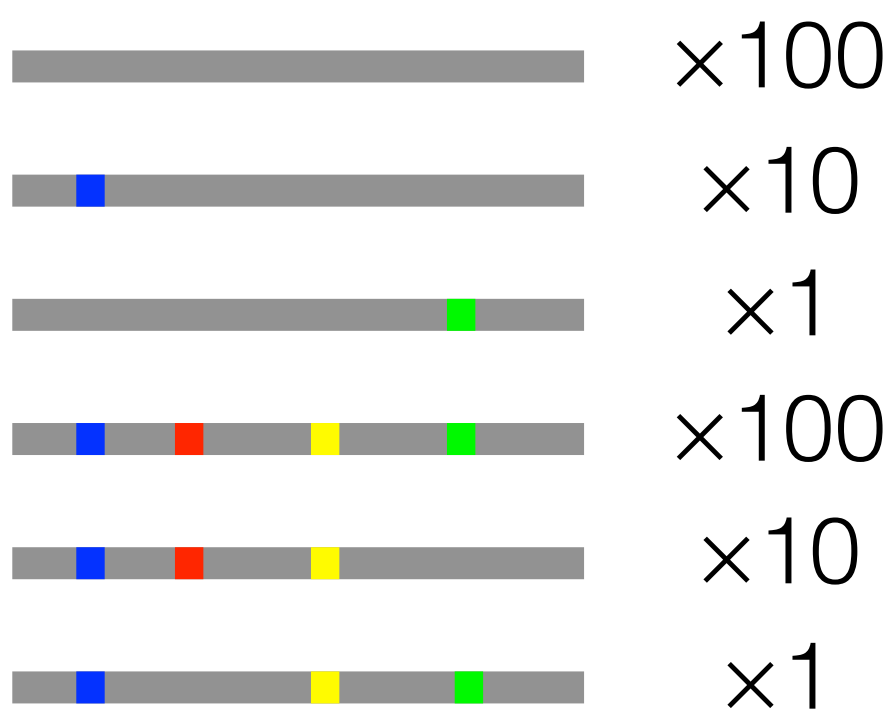
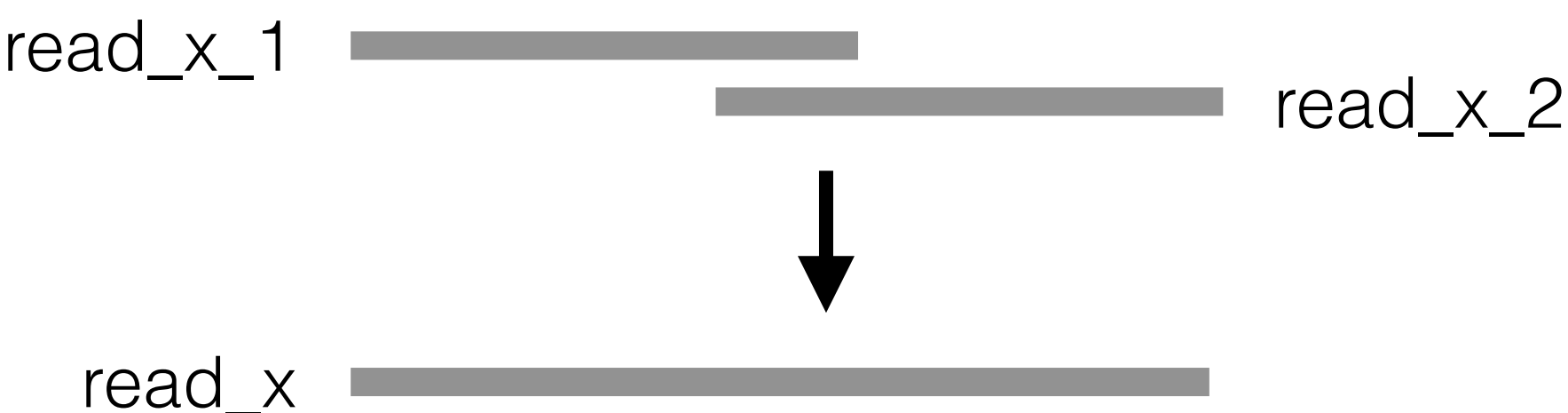


```
$ phyloscanner_make_trees.py ListOfBamFiles.csv --windows 1,300,301,600,...  
$ phyloscanner_analyse_trees.R TreeFiles OutputString ChoiceOfHostStateReconstruction
```



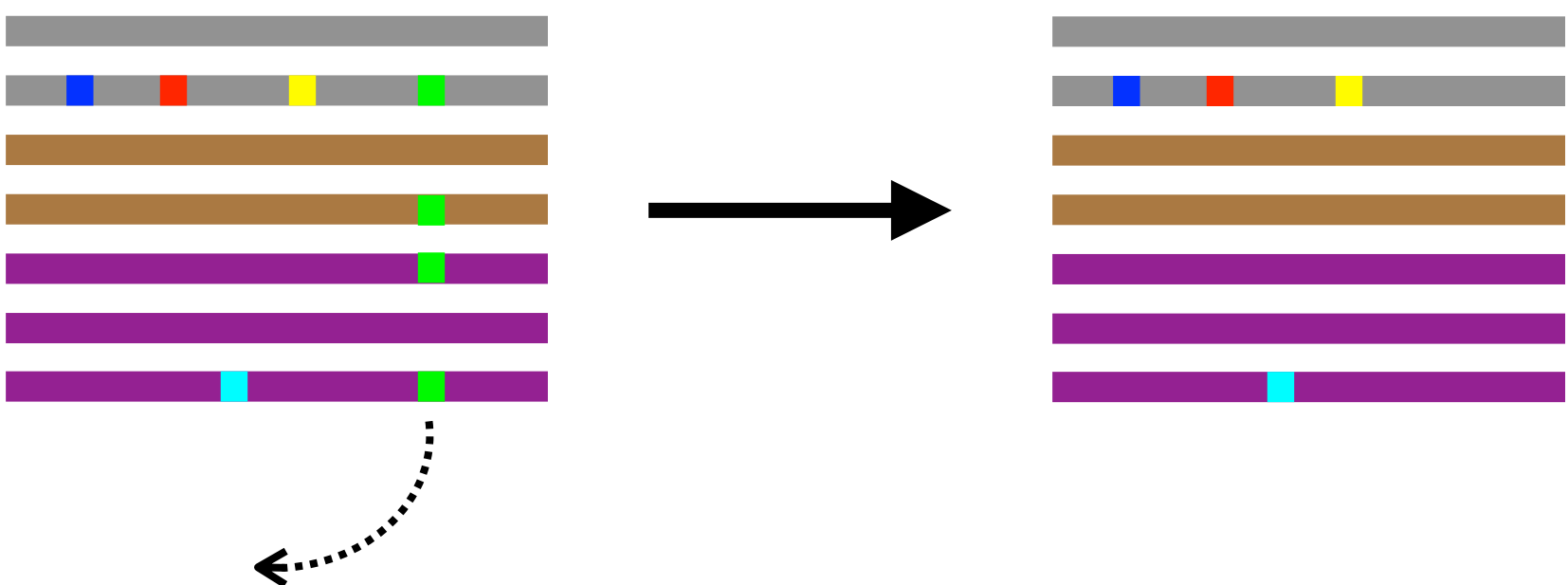
# Options

Merge overlapping paired reads into longer reads.



Similarity- and frequency-based read merging, for speed.

Excise specified positions (e.g. sites under selection).



# Options

- Include known references with the reads
- Trim and/or discard low-quality reads
- Minimum read count
- Pass any **RAxML** options, e.g. bootstraps, model specs
- Choice of ancestral host-state reconstruction algorithms & parameters
- Transmission inference parameters, e.g. distance thresholds, distance normalisation over the genome

Trivially parallelisable: split the whole genome into windows, run each window as a separate job on your cluster (in addition to multithreading **RAxML**).