

Inferring Viral Transmission Using Closely Related Virus Genomes

Mauriana Pesaresi Seminar Series 2021/2022

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Fri 8 April 2022



Introduction

Problem Statement

What

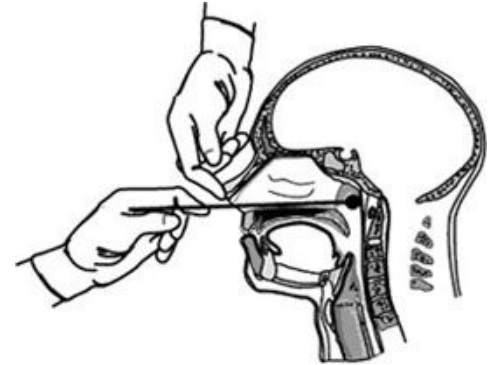
- Determine who infects whom within a household

Why

- Better prepare for imminent viruses
- Understand seasonality of viruses
- Protect the vulnerable

How

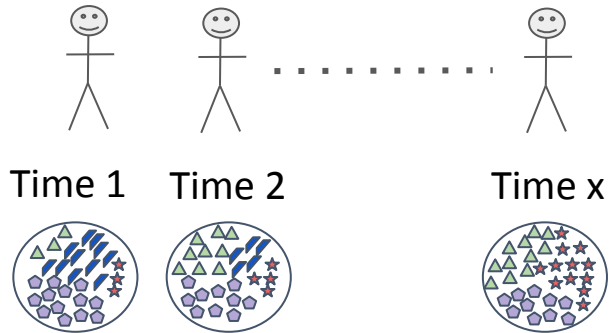
- Sample each individual often
 - nasopharyngeal swabs twice a week
- Perform an all vs all comparison of virus genomes
 - Subtext: genome comparison is not all vs all



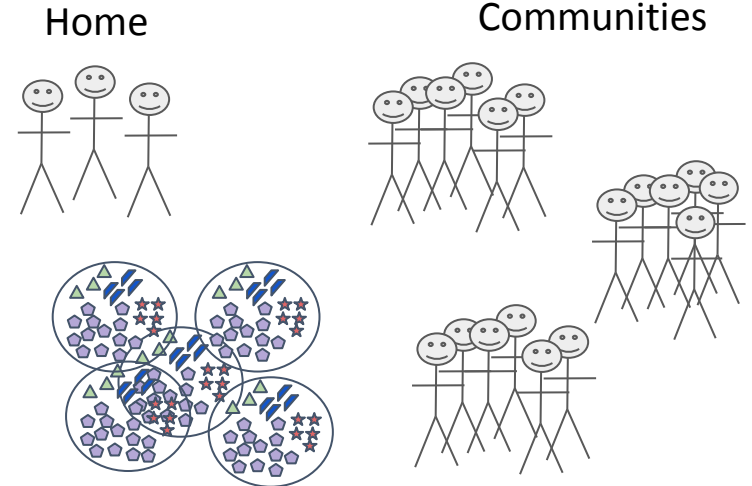
Virus Quasispecies

A viral infection is composed of multiple variants that change over time

within an individual



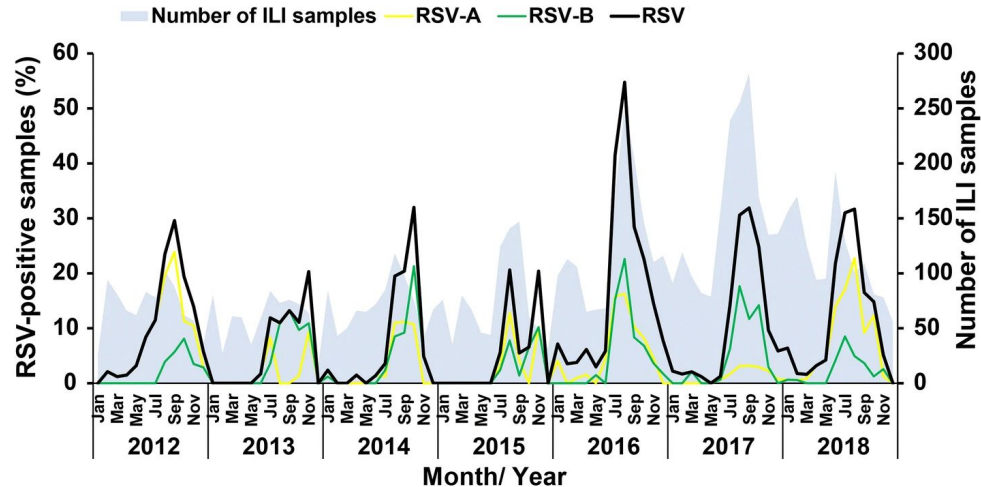
between individuals



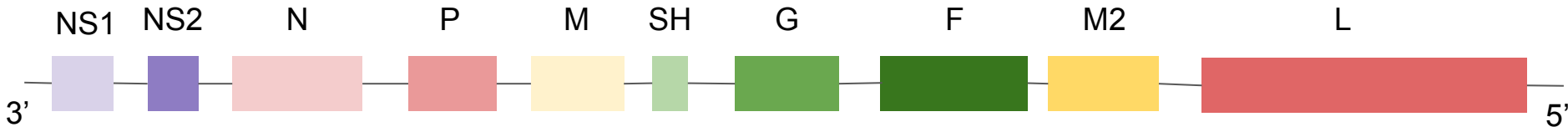
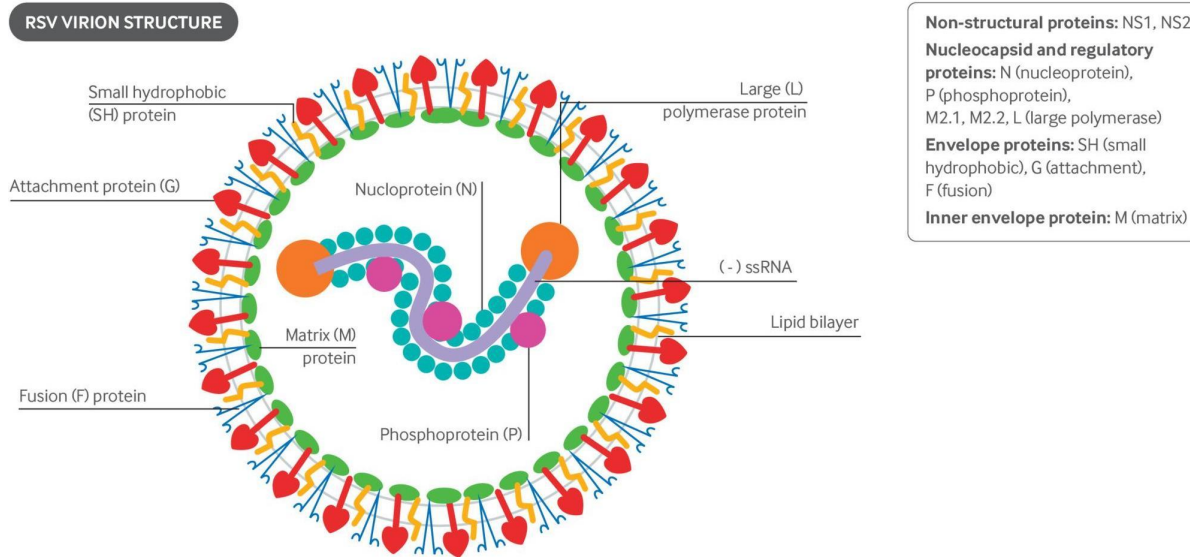
Seasonality of Viruses

Seasonality

- Predictable pattern in infections based on annual changes in the environment
- Expected in viruses transmitted through respiratory droplets



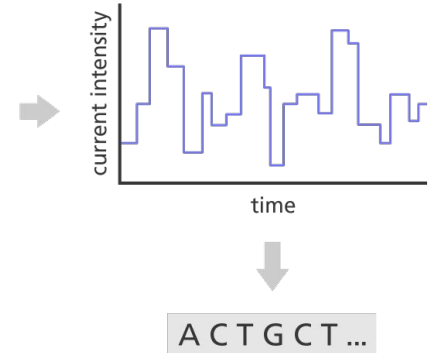
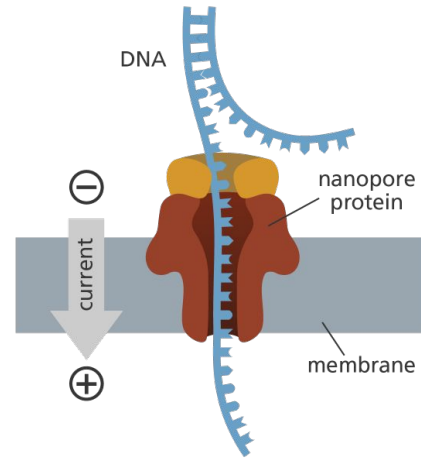
Viruses (RSV)



What is DNA data?

DNA is a chain (strand) of molecules

- Adenine (A)
- Guanine (G)
- Thymine (T)
- Cytosine (C)



DNA data is extracted through a process called **sequencing**

- modern methods run a DNA strand through a pore

Result is a **string** of A, T, C and Gs

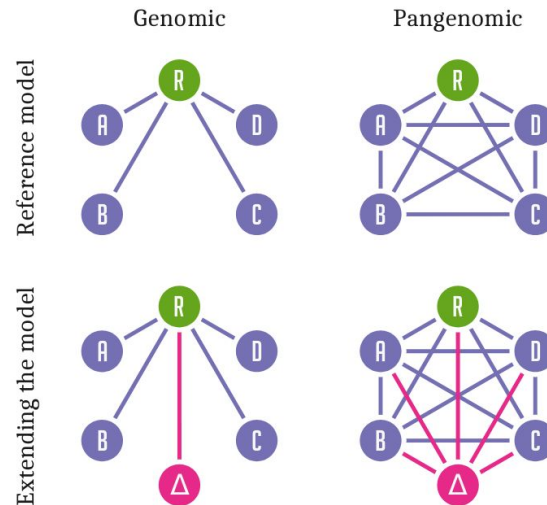
Pan-genome

A collection of many genomes

- Pan (many) genome (genetic material of an organism)

Map newly sequenced samples to a pangenome to determine similarity

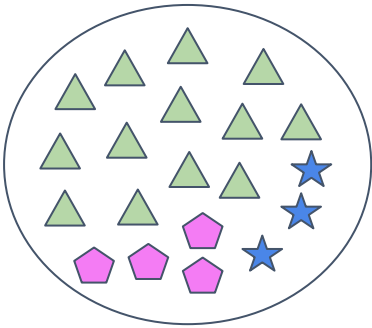
- Map: find the closest genome to a newly sampled genome



Comparing Genomes

Linear Reference Genome

A reference genome is a consensus of the most abundant sequence and is traditionally linear



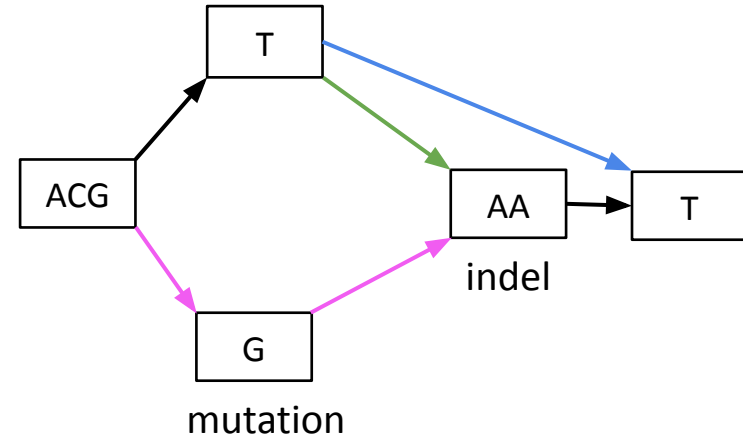
	1	2	3	4	5	6	7
Variant 1	A	C	G	T	A	A	T
Variant 2	A	C	G	T	-	-	T
Variant 3	A	C	G	G	A	A	T
Consensus	A	C	G	T	A	A	T

Pangenomic (Graphical) Reference Genome

The same consensus from the previous slide

	1	2	3	4	5	6	7
Variant 1	A	C	G	T	A	A	T
Variant 2	A	C	G	T	-	-	T
Variant 3	A	C	G	G	A	A	T
Consensus	A	C	G	T	A	A	T

The three sequences presented as a graph



Genomic Sample Comparison

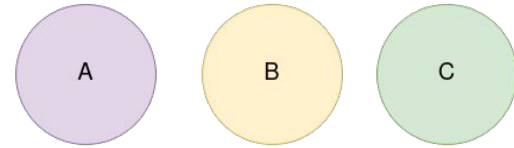
Genomic sample similarity or difference is based on how samples compare against a reference

Linear reference

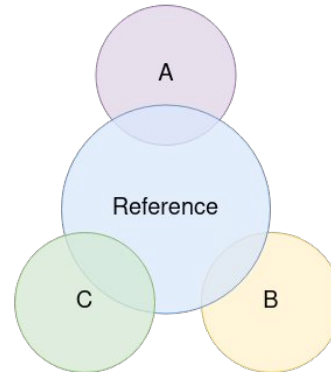
- less precise
- reference bias
- non-transitive (1 < 2 < 3 therefore 1 < 3)

Increase precision by comparing against a graphical reference

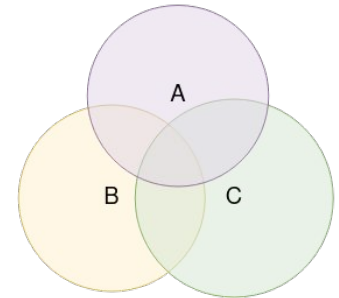
A set of samples



Linear Reference



What genome graphs attempt to do



Alignment, Mapping and Variant Calling

Align (edit distance)

- maximize similarity between two strings
 - Reward matches
 - Penalize mismatches, gaps, and gap extensions

Map

- find the **location** of a query in a text (reference)

Variant calling

- assemble reads into contigs
- map reads to a reference
- a variant is called when the number of reads with differences against the reference meets a set threshold

Align **DROWN** and **GOWN**

D	R	O	W	N
G	-	O	W	N

Align **TTAAT** and **TT**

T	T	A	A	T
T	T			
	T	T		
	T	-	-	T

Mapping Against a Linear Reference

Mapping 3 reads to a linear reference

- TAAT
- TT
- GAAT

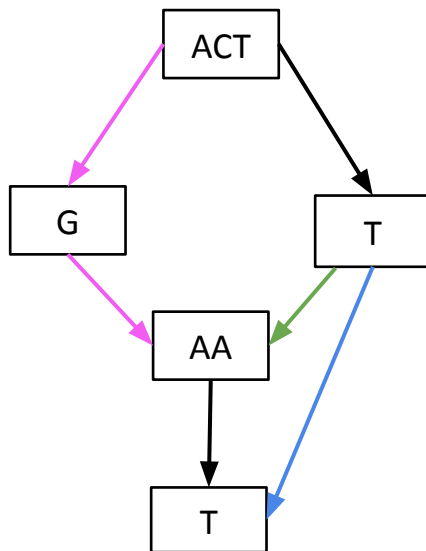
Linear reference	A	C	G	T	A	A	T
Read 1				T	A	A	T
Read 2			T	T			
Read 3				G	A	A	T

A	C	T	T	A	A	T	Variant 1
A	C	G	T	T			Variant 2
A	C	T	G	A	A	T	Variant 3

Mapping Against a Graphical Reference

Mapping 3 reads to a graphical reference

- TAAT
- TT
- GAAT



Graphical Reference						
1	A					
2	C					
3	G					
4	T	G	T	T	G	
5	A		-	A	-	A
6	A		-	A	-	A
7	T			T	T	T

Linear vs Graphical Reference Mapping

Compare the following sequences to a reference:

- TT
- TAAT
- GAAT

	1	2	3	4	5	6	7
Linear Reference	A	C	G	T	A	A	T
Read 1				T	A	A	T
Read 2			T	T			
Read 3				G	A	A	T

			Graphical Reference			
					A	1
					C	2
					T	3
T	T	G		G	T	4
A	-	A	-		A	5
A	-	A	-		A	6
T	T	T			T	7

Reference Bias

An increased likelihood of tools and methods used for read mapping to fail to identify variation and over-report variants present in the reference compared to the variants that are not present in the reference.

Effect of read-mapping biases on detecting allele-specific expression from RNA-sequencing data

Jacob F. Degner , John C. Marioni , Athma A. Pai, Joseph K. Pickrell, Everlyne Nkadori, Yoav Gilad , Jonathan K. Pritchard  [Author Notes](#)

Bioinformatics, Volume 25, Issue 24, 15 December 2009, Pages 3207–3212, <https://doi.org/10.1093/bioinformatics/btp579>

Published: 06 October 2009 [Article history](#) ▼

nature genetics

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Published: 27 April 2015

Improved genome inference in the MHC using a population reference graph

Alexander Dilthey , Charles Cox, Zamin Iqbal, Matthew R Nelson & Gil McVean 

nature biotechnology

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Published: 01 October 2018

Variation graph toolkit improves read mapping by representing genetic variation in the reference

Erik Garrison , Jouni Sirén, Adam M Novak, Glenn Hickey, Jordan M Eizenga, Eric T Dawson, William Jones, Shilpa Garg, Charles Markello, Michael F Lin, Benedict Paten & Richard Durbin 

Analysis

Coverage

How many substrings match a section of a genome

- Caused by similarity

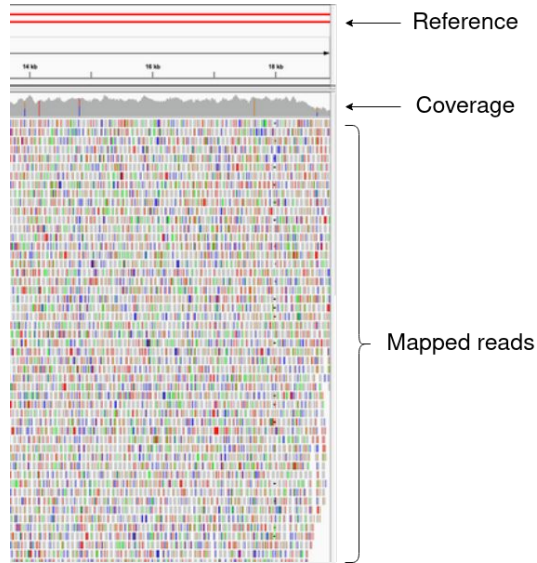
Find *aa*, *bb*, *aa*, *baa*, *aba*, *baba* in *abbaababa* exactly

0	1	2	3	4	5	6	7	8
a	b	b	a	a	b	a	b	a
						a	b	a
					b	a	b	a
			a	a				
	b	b						
		b	a	a				
				a	b	a		

Linear vs Graphical Reference Mapping

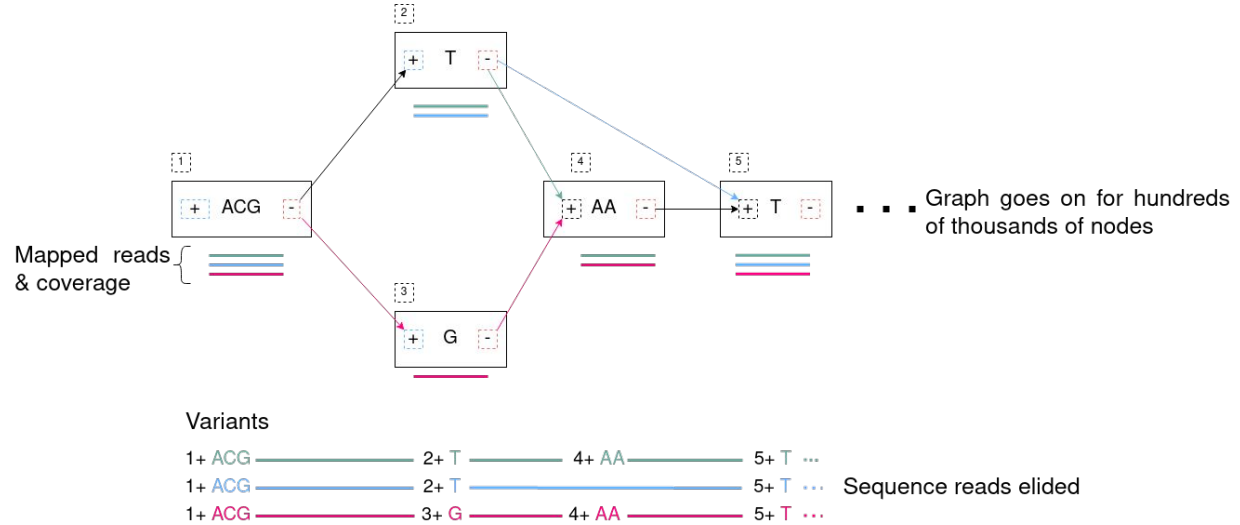
Linear Reference

Reads stack up below the loci they map against



Graphical Reference

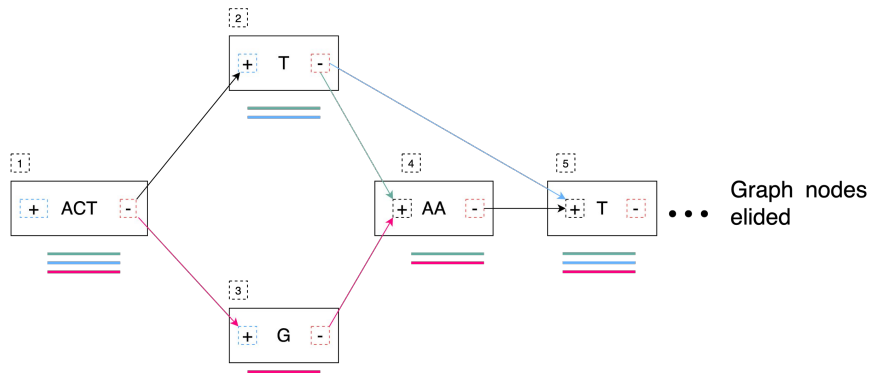
Reads stack up below the nodes whose sequences they map against



Coverage Vector and Coverage Statistics

Graphical Reference

Reads stack up below the nodes whose sequences they map against



Variants

1+ ACT ————— 2+ T ————— 4+ AA ————— 5+ T ...
 1+ ACT ————— 2+ T ————— 5+ T ...
 1+ ACT ————— 3+ G ————— 4+ AA ————— 5+ T ...

Sequence reads elided

Nodes \ Samples	X1	X2	X3	...	Xn
1	X1 ₁	X2 ₁	X3 ₁	...	Xn ₁
2	X1 ₂	X2 ₂	X3 ₂	...	Xn ₂
⋮	⋮	⋮	⋮	⋮	⋮
m	X1 _m	X2 _m	X3 _m	...	Xn _m

Pairwise Distances

Coverage Statistics

Pairwise Distances (Euclidean)

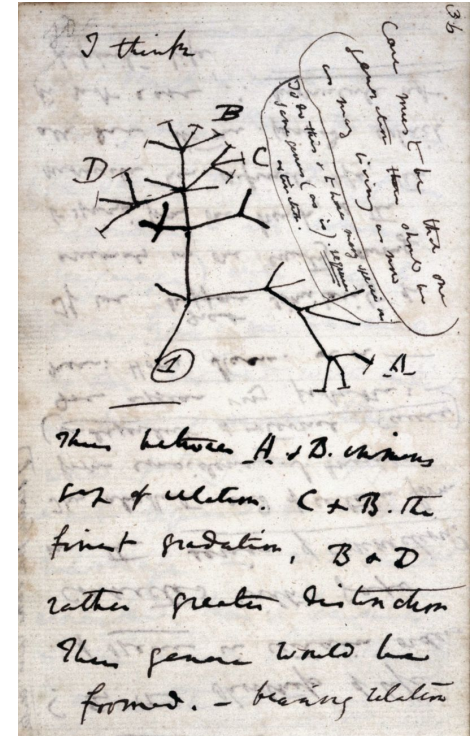
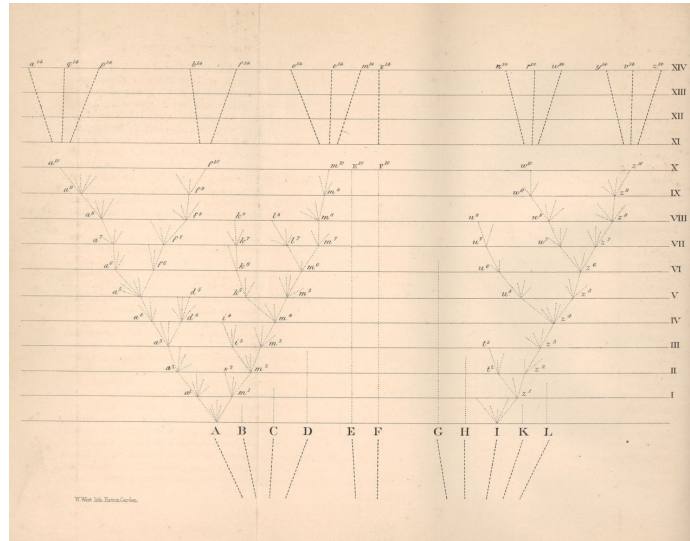
Nodes Samples	X1	X2	X3	...	Xn
1	X1 ₁	X2 ₁	X3 ₁	...	Xn ₁
2	X1 ₂	X2 ₂	X3 ₂	...	Xn ₂
⋮	⋮	⋮	⋮	⋮	⋮
m	X1 _m	X2 _m	X3 _m	...	Xn _m

$$\sqrt{(x_{1_1} - x_{1_2})^2 + (x_{2_1} - x_{2_2})^2 + \dots + (x_{n_1} - x_{n_2})^2}$$

Phylogenetics and Phylogenetic Trees

Evolutionary history and relationships between or within groups of organisms

- Analog: the tree of life



Phylogenetics and Phylogenetic Trees

Evolutionary distance

- A measure of genetic difference (mutation)

A & C are more similar

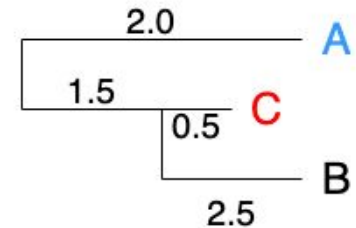
B & C are more similar

A & B are further apart from each other but closer to C

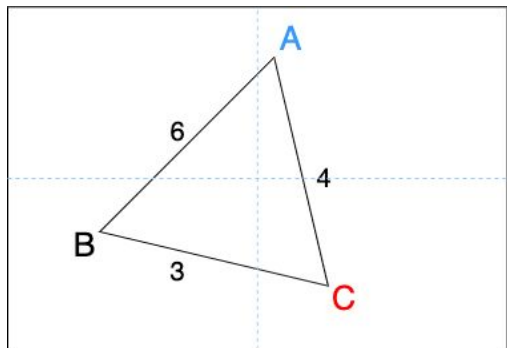
$$\text{A to C } 2.0+1.5+0.5 = 4$$

$$\text{A to B } 2.0+1.5+2.5 = 6$$

$$\text{C to B } 0.5+2.5 = 3$$

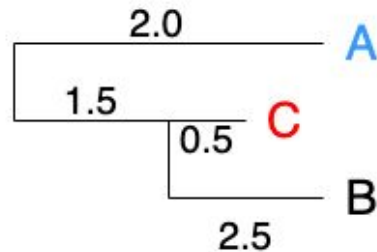


Neighbour Joining



	A	B	B
A	0	6	4
B	6	0	3
C	4	3	0

A to C $2.0+1.5+0.5 = 4$
A to B $2.0+1.5+2.5 = 6$
C to B $0.5+2.5 = 3$

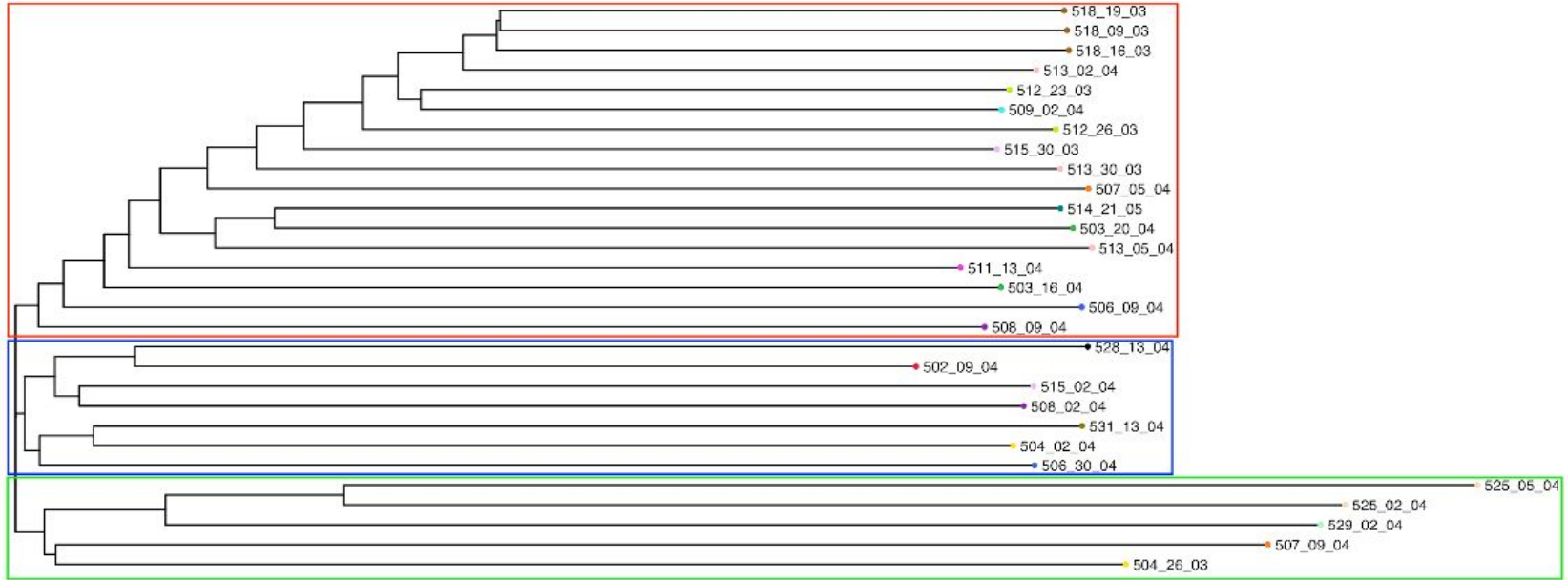


Neighbour Joining (Example)

RSV Neighbour Joining Tree

Individuals

- 502
- 503
- 504
- 506
- 507
- 508
- 509
- 511
- 512
- 513
- 514
- 515
- 518
- 525
- 528
- 529
- 531

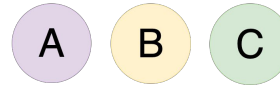


Conclusion and Further Work

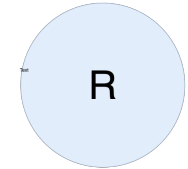
Conclusion

- A pangenomic reference can be used for sample comparison
- It is possible to perform all versus all sample comparison by comparing the coverage of reads to a graph

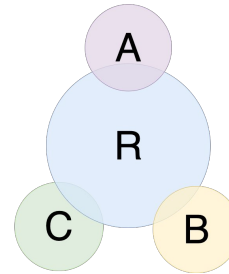
A Set of Samples



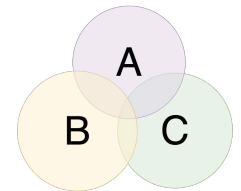
Linear Reference



Methods Based on Linear References



Methods Based on Graphical References



Conclusion

- The linear reference loses information
- A pan-genome representation is comprehensive
- By comparing new samples to a pan-genome we can
 - compare samples better
 - infer transmission with a higher degree of certainty
 - better prepare for imminent variants

Conclusion

Assumptions

- Sample relatedness implies potential transmission
- Minimum evolution (neighbour joining assumes minimum evolution)
- Bases have an equal probability of substitution (JC69 model)
- Coverage under each node is equally informative

Further work

- Extend the approach to other households
- Extend the approach to more sparse testing time periods
- Use more robust phylogenetic methods than neighbour joining

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