

Viral Genomics and Bioinformatics (Latin America and the Caribbean)



10–14 October 2022 - Virtual course

Introduction to phylogenetic methods

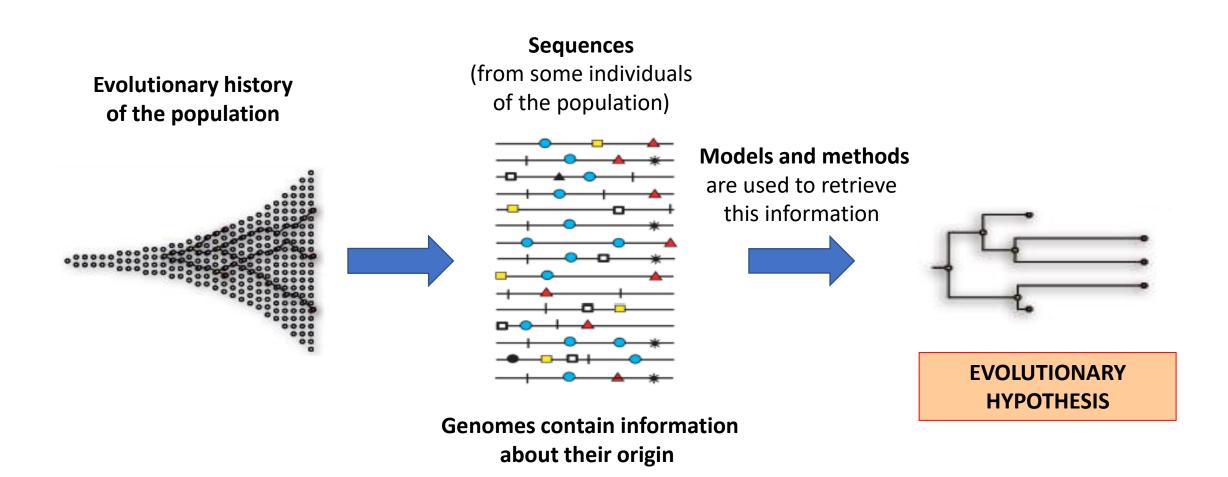
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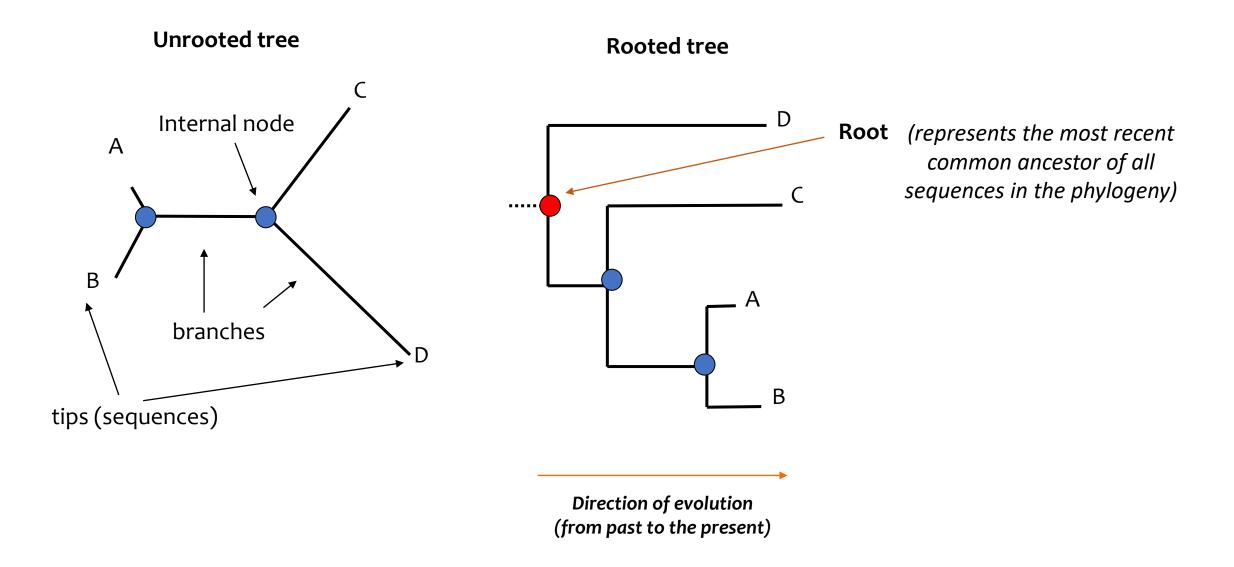
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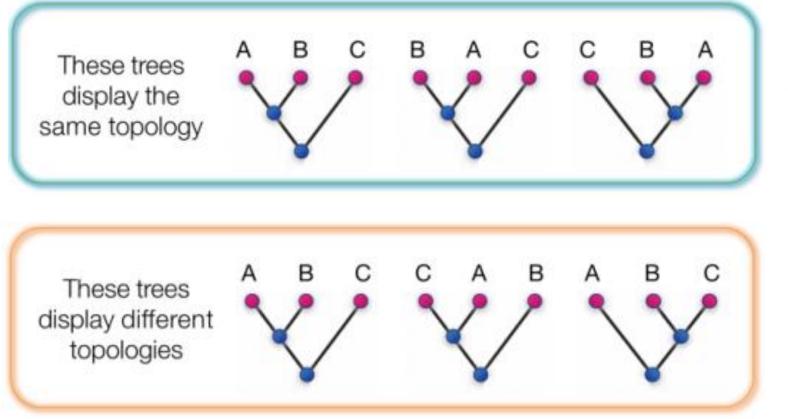




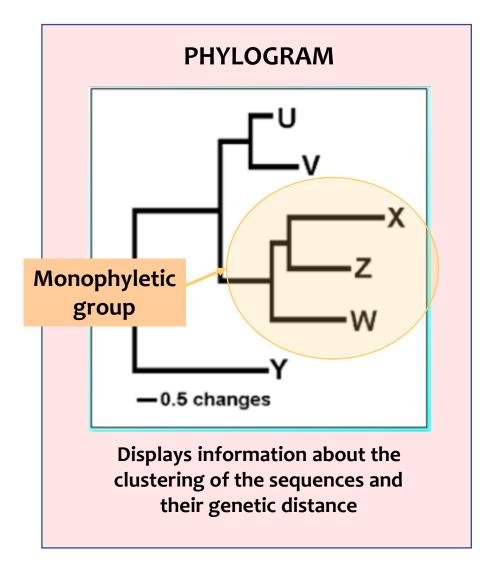
Phylogenetic analysis allows us to reconstruct the evolutionary history and study the processes that gave rise to it.



• The **topology** is the branching structure of the tree.



Branches can rotate freely over nodes



Monophyletic group: group in which all sequences of interest share a common ancestor (without any other sequences within the group).



A common source of infection or transmission chain.

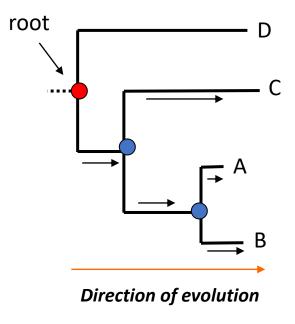
Methods for rooting phylogenetic trees

Outgroup Rooting

- ✓ The outgroup is a group of taxa whose diversification of the group of interest occurred before the internal diversification of the taxa under study (ingroup).
- ✓ The outgroup must belong to a clearly different lineage from the ingroup, but it does not have to be so divergent that it cannot be aligned unambiguously.
- \checkmark The root is set on the branch that connects the outgroup to the ingroup.

• Midpoint Rooting

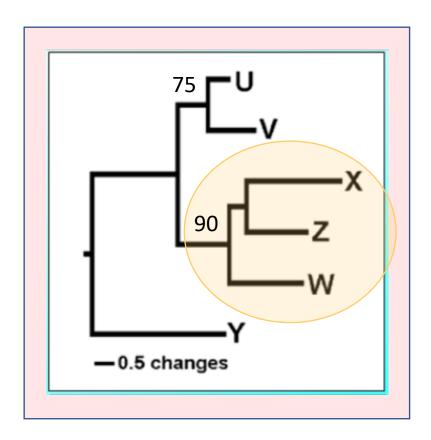
✓ The root is established at the midpoint distance between the two most divergent taxa.



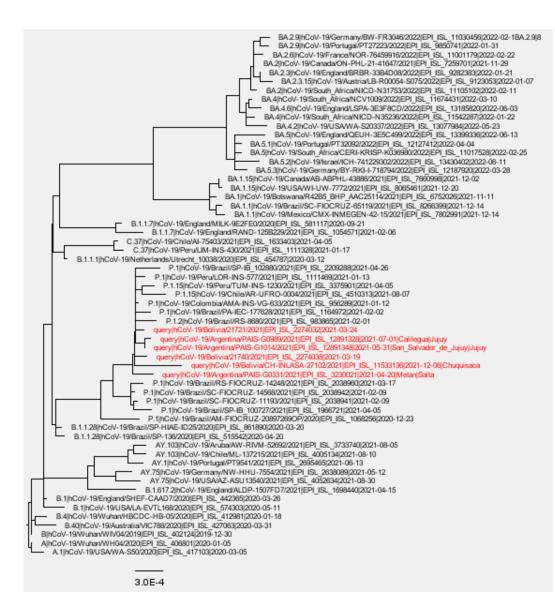
The root allows to establish the direction of the evolutionary process.

How are phylogenetic trees analysed and interpreted?

- **Root the tree** to know the direction of evolution (branch connecting the *ingroup* with the *outgroup* or midpoint).
- Analyse the groupings from the root to the terminals.
- Identify the common ancestor of the sequences of interest and analyse with which sequences they are related (with whom they share ancestors).
- Assess the **reliability of the clusters of interest** (how much evidence is there in the sequences for the formation of those clusters).
- Analyse the evolutionary relationships together with the rest of the information available (metadata): epidemiological, biological, temporal, geographical information, etc.



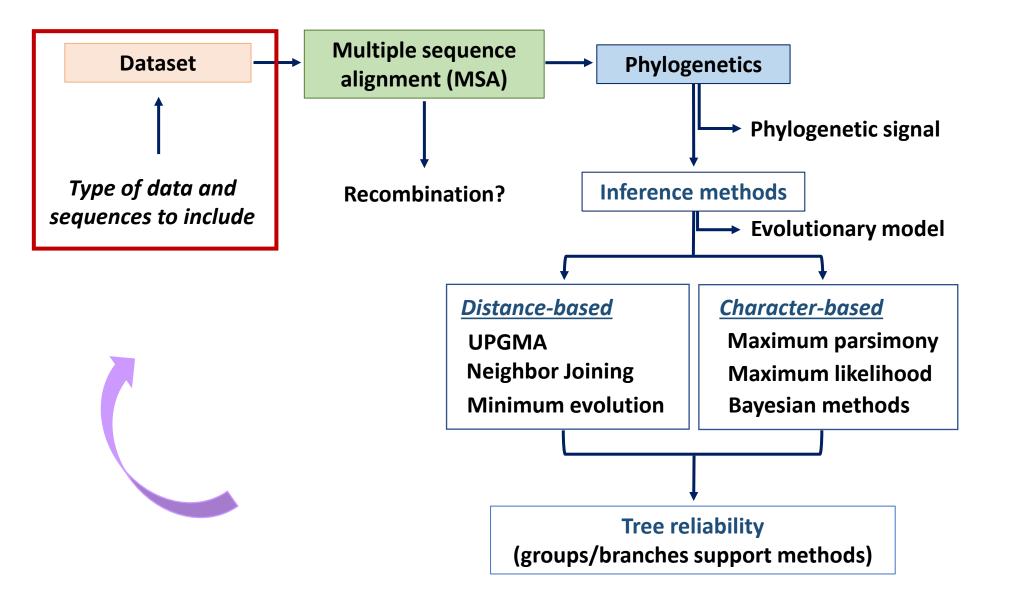
Phylogenetic analysis



- Are the sequences in red related?
- Do they belong to the same transmission chain?

...we must prepare a dataset that allows us to address this objective!

Steps for phylogenetic analysis



Sequence selection for molecular phylogenetics

- What sequences to include?
 - ✓ Sequences under study
 - ✓ Reference sequences
 - ✓ Outgroups
 - Other sequences (according to the objective): sequences of the organism under study that circulates in the same geographical region or during the same period, sequences with a high % identity or that show specific characteristics of interest (host, virulence, clinical behavior, etc).

The entire dataset has an impact on the results and the conclusions reached

• Type of data

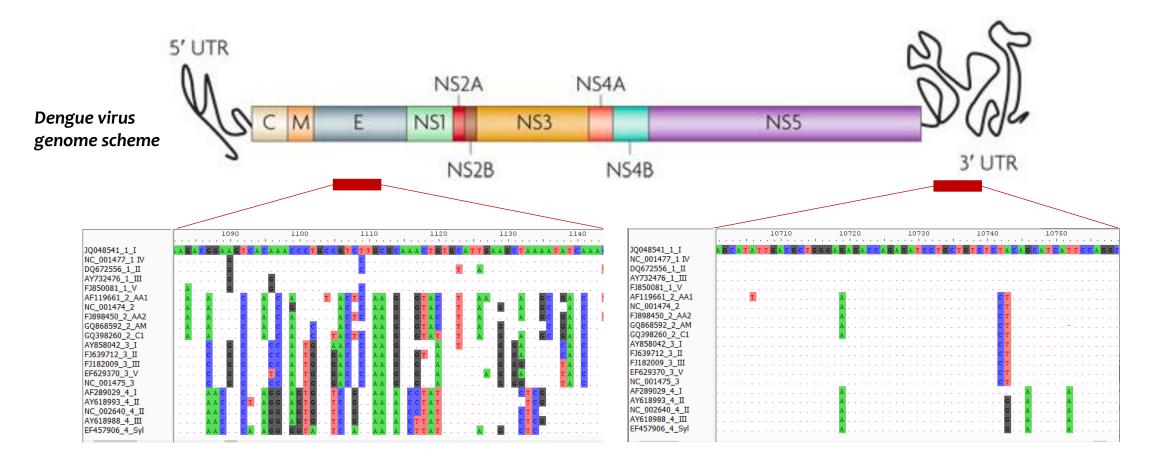
- ✓ Nucleotide sequences
- ✓ Amino acid sequences
- ✓ SNPs

The sequences to include and the type of data to be analyzed always depend on the objective.

Type of data: Sequences

- ✓ Nucleotides or amino acids.
- Complete genomes or partial sequences (which genetic region?).

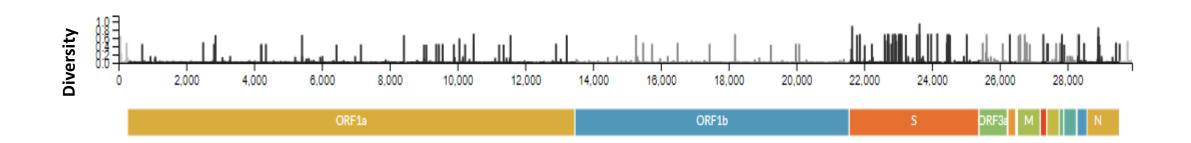
Adequate diversity for the taxonomic range under study (family, species, genotype, subgenotype, etc.).



Type of data: Sequences

- ✓ Nucleotides or amino acids.
- Complete genomes or partial sequences (which genetic region?).

Adequate diversity for the taxonomic range under study (family, species, genotype, subgenotype, etc.).



In the case of SARS-CoV-2, the complete genome is usually used

Databases



GenBank National Center for Biotechnology Information http://www.ncbi.nlm.nih.gov



International Nucleotide Sequence Database Colaboration

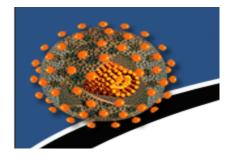
European Nucleotide Archive

European Nucleotide Archive <u>https://www.ebi.ac.uk/ena</u>



http://www.ddbj.nig.ac.jp/Welcome.html.en





HIV DATABASES

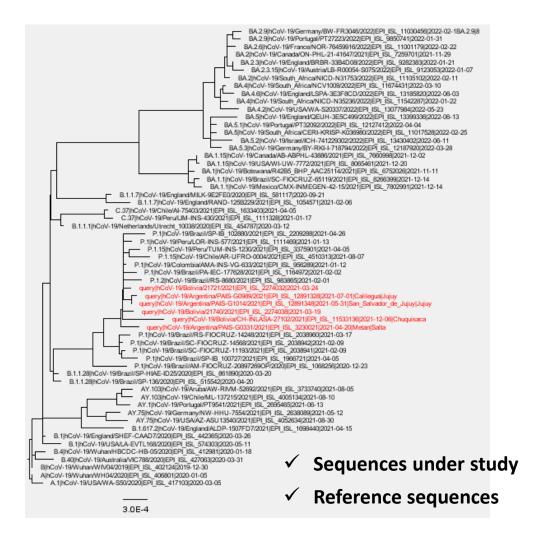


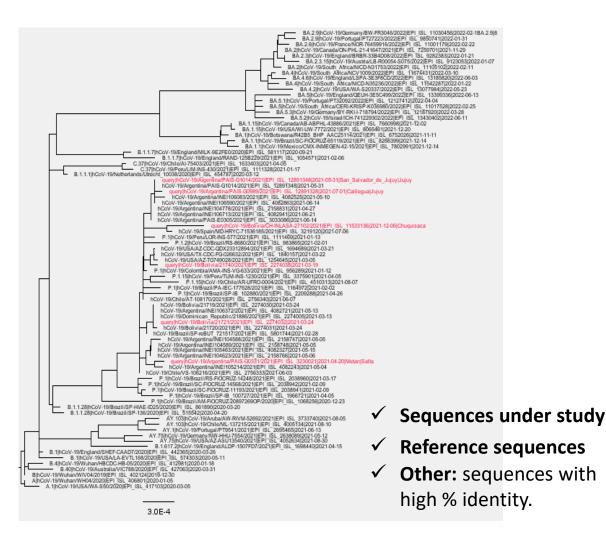
Global Initiative on Sharing All Influenza Data



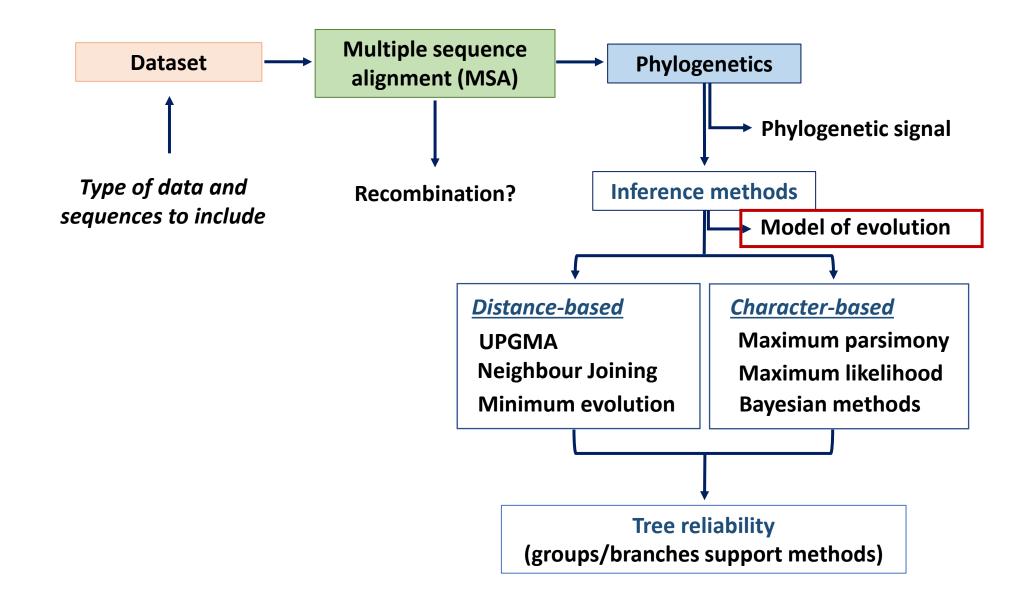
Phylogenetic analysis

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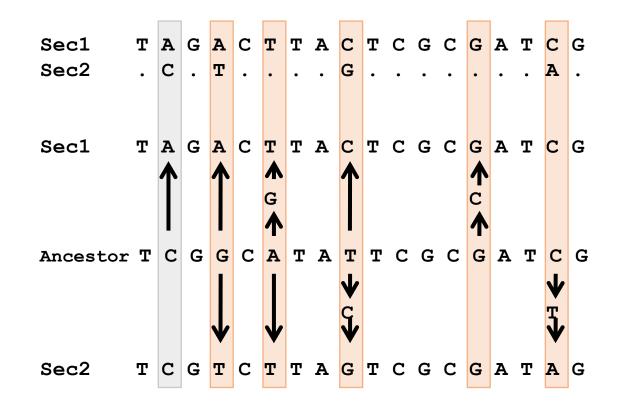




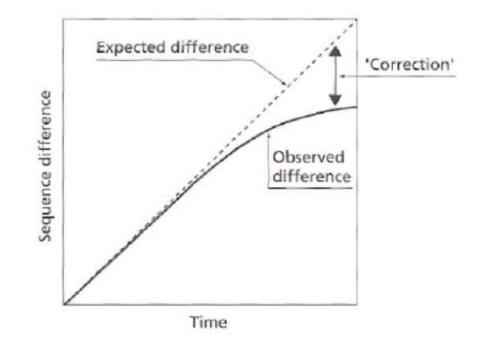
Steps for phylogenetic analysis



• They are mathematical formulations that model the type and rate of change per site.



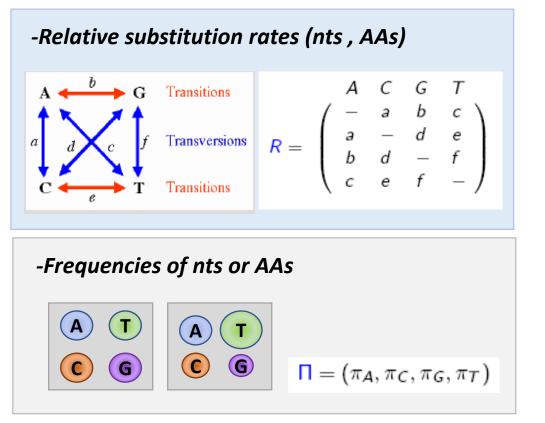
Throughout the evolutionary history of sequences, some positions may have changed multiple times.



They allow distances to be "corrected" for mutations or unobserved changes

• They are mathematical formulations that model the type and rate of change per site.

Model Parameters:

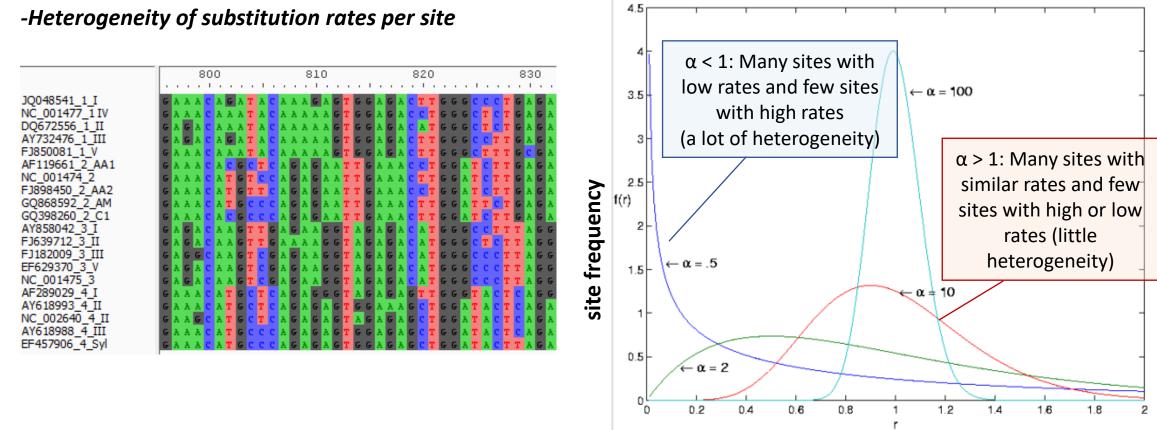


Model	df	Explanation	Code			
JC or JC69	0	Equal substitution rates and equal base frequencies (Jukes and Cantor, 1969).				
F81	3	Equal rates but unequal base freq. (Felsenstein, 1981).				
K80 or K2P	1	Unequal transition/transversion rates and equal base freq. (Kimura, 1980).	010010			
HKY or HKY85	4	Unequal transition/transversion rates and unequal base freq. (Hasegawa, Kishino and Yano, 1985				
TN or TN93	5	Like HKY but unequal purine/pyrimidine rates (Tamura and Nei, 1993).				
TNe	2	Like TN but equal base freq.				
K81 or K3P	2	Three substitution types model and equal base freq. (Kimura, 1981).				
K81u	5	Like K81 but unequal base freq.				
TPM2	2	AC=AT, AG=CT, CG=GT and equal base freq.				
TPM2u	5	Like TPM2 but unequal base freq.				
TPM3	2	AC=CG, AG=CT, AT=GT and equal base freq.				
TPM3u	5	Like TPM3 but unequal base freq.				
TIM	6	Transition model, AC=GT, AT=CG and unequal base freq.				
TIMe	3	Like TIM but equal base freq.				
TIM2	6	AC=AT, CG=GT and unequal base freq.				
TIM2e	3	Like TIM2 but equal base freq.				
TIM3	6	AC=CG, AT=GT and unequal base freq.	012032			
TIM3e	3	Like TIM3 but equal base freq.				
TVM	7	Transversion model, AG=CT and unequal base freq.				
TVMe	4	Like TVM but equal base freq.				
SYM	5	Symmetric model with unequal rates but equal base freq. (Zharkikh, 1994).				
GTR	8	General time reversible model with unequal rates and unequal base freq. (Tavare, 1986).				

• Each site is assumed to evolve independently (nts, AAs, codons, etc.)

• They are mathematical formulations that model the type and rate of change per site.

Model Parameters:



Gamma Distribution

Relative substitution rate

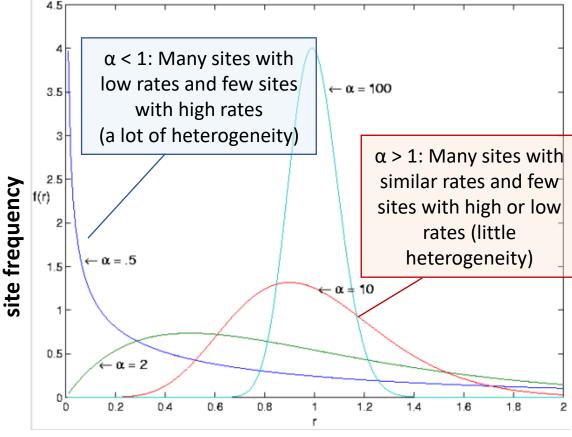
• They are mathematical formulations that model the type and rate of change per site.

Model Parameters:

-Heterogeneity of substitution rates per site

- +**G**: Gamma distribution (parameter α).
- +I: proportion of Invariant sites (p-inv).
- +G+I: Gamma distribution plus invariant sites.
- **+R:** FreeRate model (generalization of the Gamma distribution assumption).

+R+I: FreeRate model plus invariant sites.



Gamma Distribution

• **Programs:** ModelFinder (IQ-TREE), jModelTest, ProtTest, others.

Relative substitution rate

Corrected Akaike Information Criterion: GTR+F+R3

Best-fit model: GTR+F+R2 chosen according to BIC

Bayesian Information Criterion:

Example: Selection of the evolutionary model with IQ-TREE (ModelFinder)

Mode	lFinder will te	est 286 DNA	models	s (sample size	e: 29796)	
No.	Model	-LnL	df	AIC	AICc	BIC
1	JC	46744.121	149	93786.242	93787.749	95023.259
2	JC+I	46642.012	150	93584.023	93585.552	94829.343
3	JC+G4	46675.504	150	93651.008	93652.536	94896.327
4	JC+I+G4	46650.941	151	93603.882	93605.430	94857.503
5	JC+R2	46719.714	151	93741.428	93742.977	94995.050
6	JC+R3	46669.794	153	93645.587	93647.177	94915.813
7	JC+R4	46639.294	155	93588.587	93590.219	94875.417
8	JC+R5	46634.671	157	93583.342	93585.016	94886.777
14	F81+F	45807.542	152	91919.084	91920.653	93181.008
15	F81+F+I	45710.258	153	91726.516	91728.106	92996.742
16	F81+F+G4	45740.064	153	91786.129	91787.719	93056.355
17	F81+F+I+G4	45716.981	154	91741.961	91743.572	93020.489
())					
274	GTR+F	45588.826	157	91491.652	91493.326	92795.086
275	GTR+F+I	45499.127	158	91314.253	91315.948	92625.989
276	GTR+F+G4	45523.194	158	91362.388	91364.083	92674.124
277	GTR+F+I+G4	45502.417	159	91322.834	91324.551	92642.873
278	GTR+F+R2	45492.606	159	91303.212	91304.929	92623.251
279	GTR+F+R3	45486.942	161	91295.885	91297.645	92632.528
Akai	ke Information	Criterion:		GTR+F+R3		

C AICC BIC 786.242 93787.749 95023.259

selected

model

- <u>The "fit" of the data (alignment) to several possible models is</u> <u>evaluated</u>, with parameter optimization.
- <u>A criterion (AIC or BIC) is applied to select the best model</u>: good fit (awarded) vs number of parameters (penalized).
- Complex models (more parameters) are selected only if their fit to the data is substantially better than simpler models.

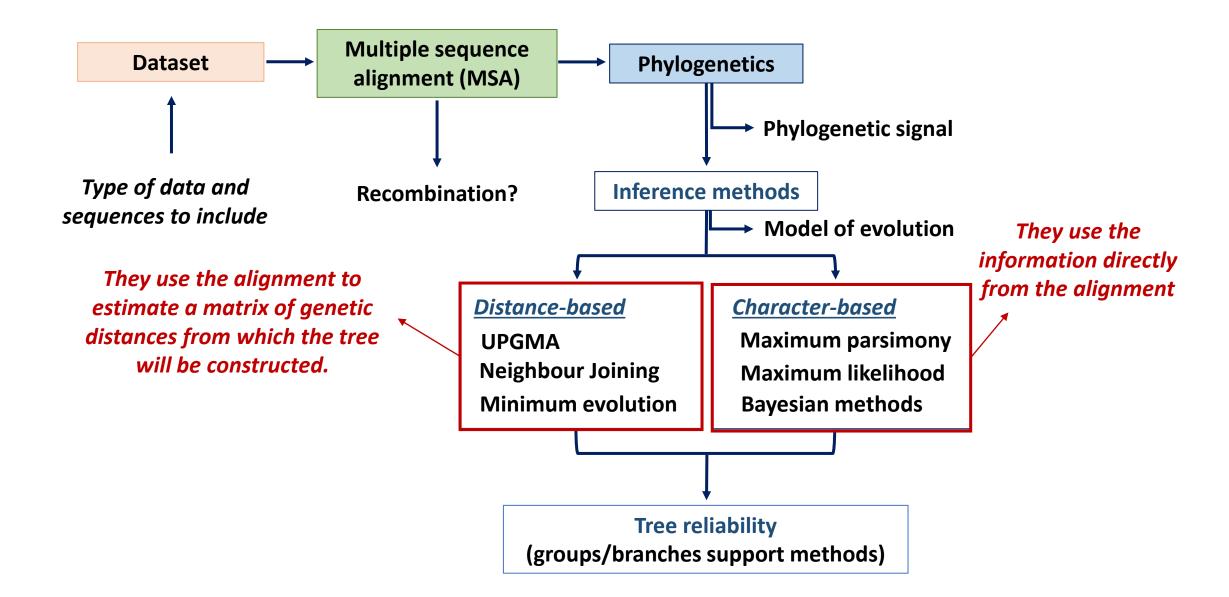
The model must be estimated for <u>each dataset</u> (There is no model that is specific to a gene, genome or organism...)

Rate parameters: A-C: 0.43193 A-G: 1.36320 A-T: 0.18530 C-G: 0.34865 C-T: 3.18238 G-T: 1.00000 Base frequencies: A: 0.299 C: 0.183 G: 0.196 T: 0.322 Site proportion and rates: (0.509,0.266) (0.491,1.762)

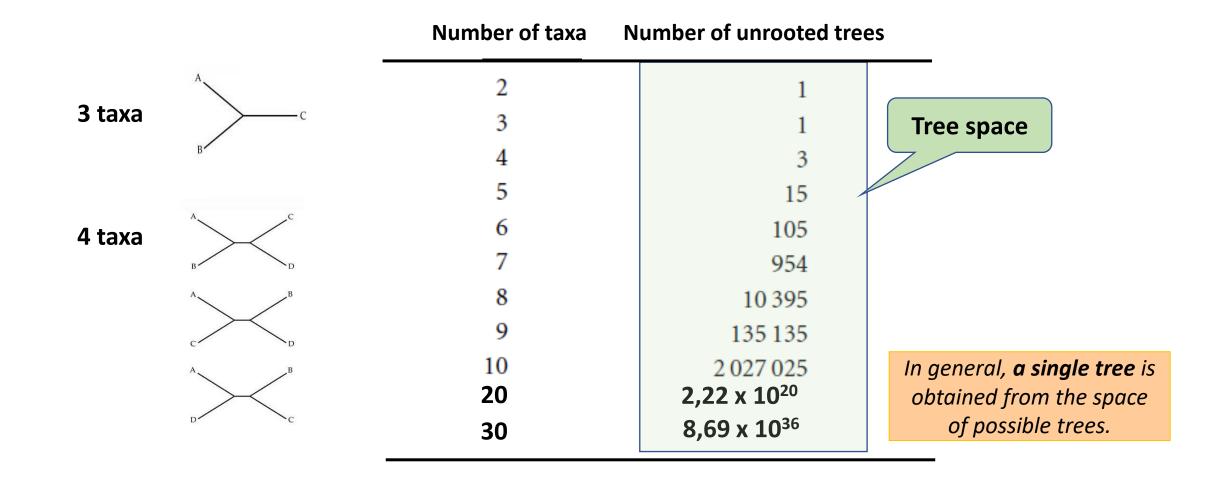
GTR+F+R2



Parameters of the selected model

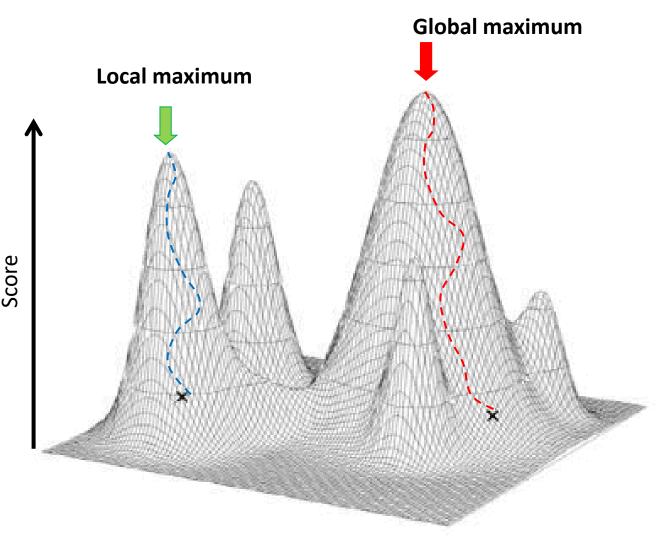


Tree space



Phylogenetic methods select one tree from the several millions/billions of possibilities.

- Distance-based: Tree building from genetic distance.
 - Resume the data of the sequences into numbers: the "distance" to the other taxa.
 - The clustering methods group the taxa from the nearest to the farthest giving only one result (Neighbour Joining).
 - The methods that follow a criterion, evaluate alternative hypotheses and select the one that minimizes the distances between taxa.
- Character-based: Tree building from character information.
 - They evaluate alternative hypotheses (tree searching) and choose which ones (trees) meet the optimality criterion (Maximum likelihood, Maximum parsimony).
 - They explore the tree space and save some alternative hypotheses that will be summarized in one tree (Bayesian methods).



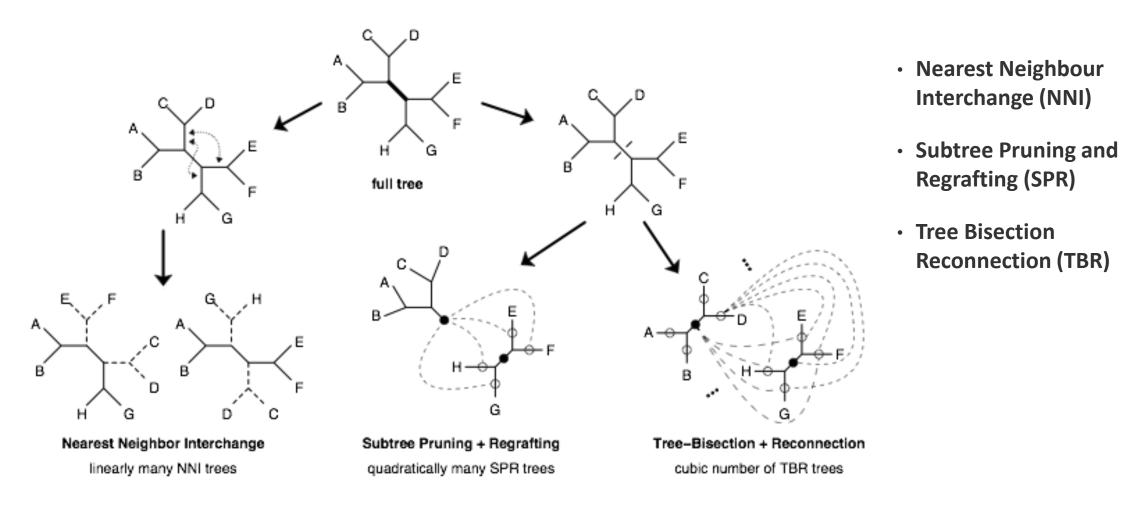
EXPLORATION OF TREE SPACE BY HEURISTIC SEARCH

- First, the initial tree (NJ or other) is built.
- The tree is evaluated under one criterion (e.g., Maximum Likelihood).
- The initial tree is modified (e.g., by rearrangement of branches) giving new trees that are also evaluated under the criterion.
- If any of the new trees is better than the initial, the first is discarded and the best one is retained.
- The process is repeated until no better trees are found under the criterion.

Methods of phylogenetic analysis

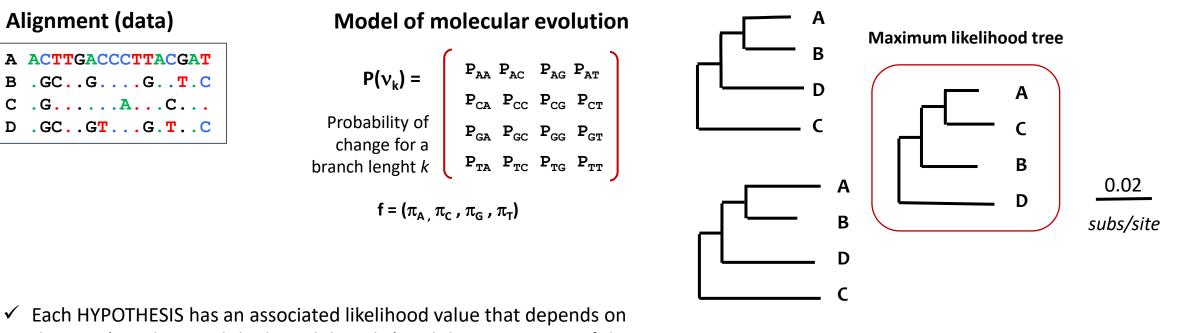
EXPLORATION OF TREE SPACE BY HEURISTIC SEARCH

• Branch rearrangements (changes in topology):



L= L1 x L2 x L3 x ... Ln (site)

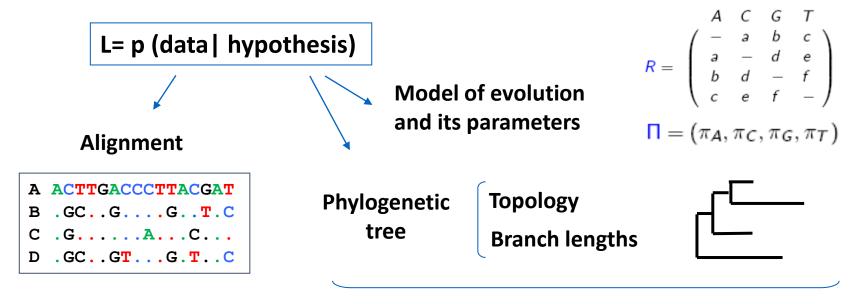
Criterion: Tree that maximizes the probability of alignment (Likelihood= P(D|H)).



- ✓ Each HYPOTHESIS has an associated likelihood value that depends on the tree (topology and the branch lengths) and the parameters of the model of evolution.
- \checkmark The hypothesis that meets the criterion of the maximum likelihood is selected.
- Programs: IQ-TREE, RAxML, PhyML, others...

Methods of phylogenetic analysis

✓ We look for the topology and the set of parameters (hypothesis) that maximize the probability of the data.

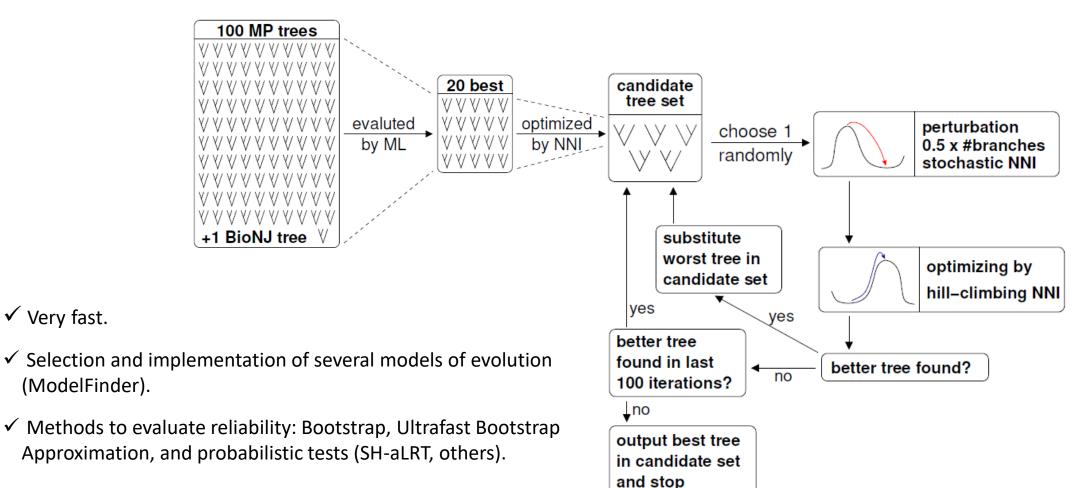


Multidimensional optimization!

(cannot be solved in one go)

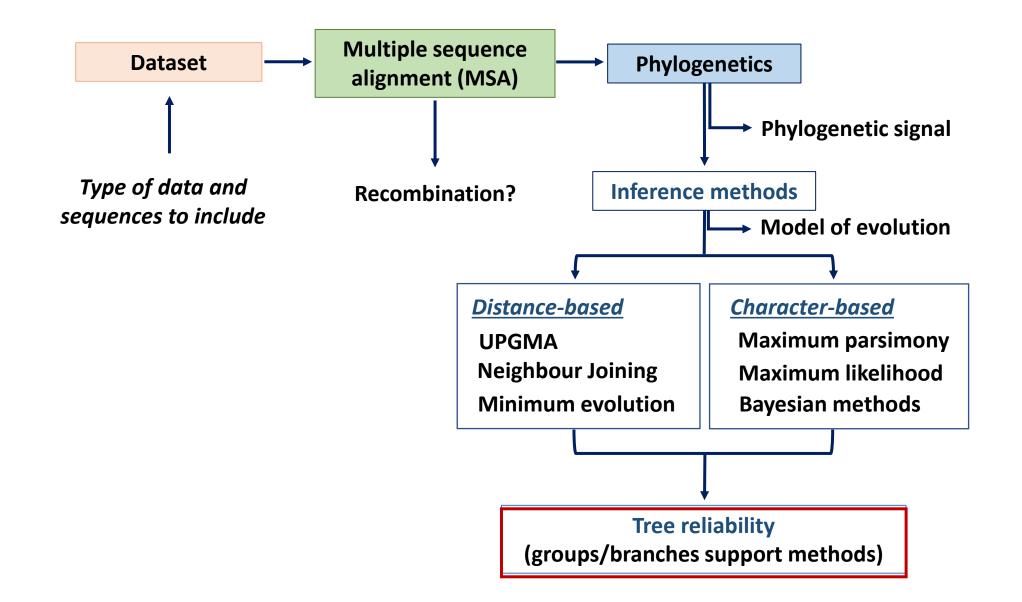
IQ-TREE

✓ Very fast.



MP/BioNJ + randomization / permutation + fastNNI

Steps for phylogenetic analysis

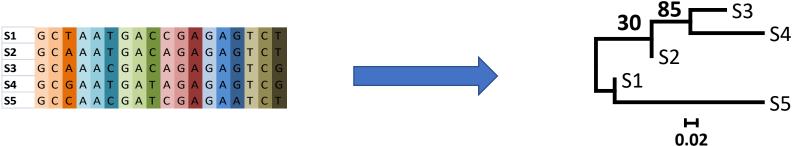


- The methods to evaluate the reliability allow us to know **how much evidence** exists in the dataset about the observed clusters, but they do NOT allow to establish if the groups exist in nature.
- Phylogenetic trees show reliable clusters in some areas and unreliable in others (higher and lower resolution spaces).
- Methods:

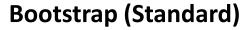
Resampling methods: Bootstrap (Standard), Rapid Bootstrap, Ultrafast Bootstrap approximation (UFB). **Probabilistic approaches:** Approximate Likelihood Ratio Test (aLRT), Shimodaira-Hasegawa-aLRT (SH-aLRT).

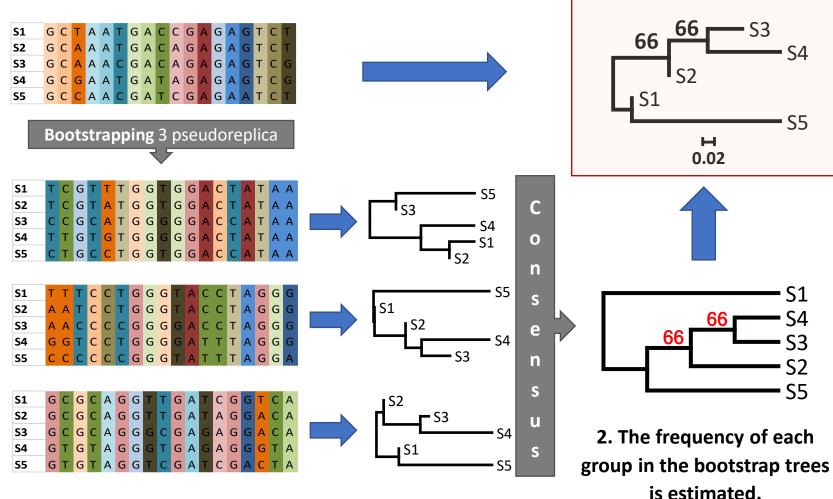
• The support values are usually indicated on the branches of the trees and there are recommendations on values considered "reliable" for each method (Bootstrap > 70-80%; UFB > 90-95%; SH-aLRT > 80%).

Example:



Tree reliability



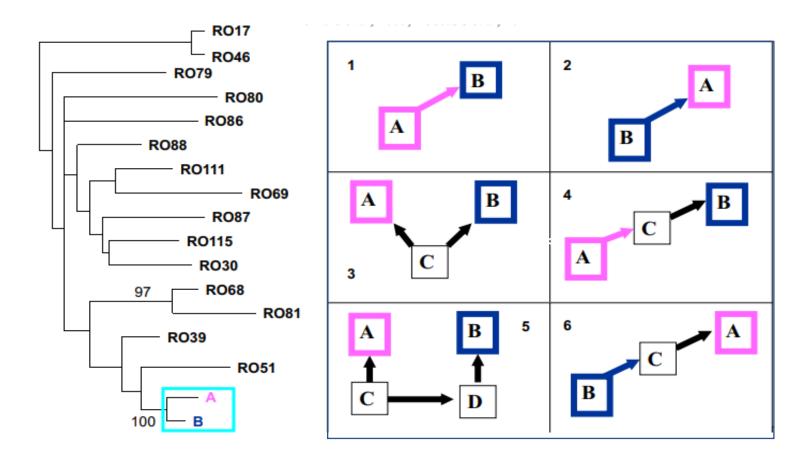


3. The bootstrap values are translated into the phylogenetic tree obtained from the actual data.

- Groups showing Bootstrap values > 70-80% are considered supported.
- It can be applied to different inference methods (MP, NJ, ML) and is implemented in several programs.
- ✓ It is very slow (demanding) and conservative.

1. The alignment is sampled repeatedly (100-10000) and for each replica, a phylogenetic tree is obtained.

A common ancestor is sought, however, there are several possible scenarios for the same topology:



Phylogenetic trees should always be interpreted in the context of all available information.

Avoid overinterpreting trees!

Other analyses using phylogenetic inference

- Phylogenetic signal.
- Formal hypothesis testing on topologies.
- Selective pressure studies.
- Sequence ancestral reconstruction.
- Exploration of the temporal signal \rightarrow Phylodynamic analyses

References

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