

Tree-thinking and basic approaches to building phylogenies

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Nothing in biology makes sense except in the light of evolution.

- Theodosius Dobzhansky, 1973

Nothing in evolution makes sense except when seen in the light of phylogeny.

- Jay Savage, 1997

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The Tree of Life

Three billion years the Tree has grown From replicators' first seed sown To branches rich with progeny: The wonder of phylogeny.

excerpt from the poem "<u>The Tree of Life</u>" by David Maddison





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CCO

nodes can be rotated without changing the relationships of the descendant branches

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a node represents the **most-recent common ancestor** (MRCA) of the descendant taxa

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this tree shows that humans and chimpanzees are more closely related to each other than either is to gorillas

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gorillas *are sister to* humans & chimps

gorillas *are NOT basal* to chimps & humans

CCO





the term basal refers to something that is "closer to the base", so please never use this for extant taxa

the 10 Mya fossil Chororapithecus abyssinicus can be called a basal ape (Sewa et al 2007)

Groups in a Phylogeny



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Monophyly

a monophyletic group includes an ancestor and all of its descendants



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Paraphyly

a paraphyletic group includes an ancestor and a subset of its descendants



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Polyphyly

a polyphyletic group includes a set of taxa, but not their common ancestor



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Phylogeny: Branch Lengths



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Phylogeny: Branch Lengths

can represent the amount of **genetic difference**

0.03 base substitutions/site

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Phylogeny: Branch Lengths

can represent the **duration of time** between nodes



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Phylogeny: Rooting

a tree can be **rooted** to show the direction or relative timing of divergence

0.03 base substitutions/site

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Phylogeny: Rooting

a tree can be **unrooted** showing only relationships among lineages



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Resolved Phylogeny

a tree can be **bifurcating** or binary when all nodes split into only 2 descendants (such a tree is also called "resolved")

fully-resolved tree



fully-resolved tree



Unresolved Phylogeny

polytomies allow us to represent unresolved nodes



star tree

partially-resolved tree



Representing Trees

((A,D),(C,B));

we can represent trees using <u>Newick</u> format, which uses sets of nested parentheses



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Representing Trees

((A:0.5,D:0.3):1.1,(C:0.1,B:0.2):0.3);

this format also accommodates branch information



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Phylogenetics & Epidemiology



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https://nextstrain.org/ncov/gisaid/global

Inferring Phylogenies

to estimate a phylogeny, we need to start with data

the data are observations of character states for a set of taxa

taxa	character 1	character 2	character 3
T1	pointed	blue	present
T2	pointed	blue	present
Т3	round	blue	absent
T4	round	black	absent

a column in the matrix is a character

the form that character takes is its **state**

Inferring Phylogenies

to estimate a phylogeny, we need to start with data

the data are observations of character states for a set of taxa

taxa	character 1	character 2	character 3
T1	Α	Τ	С
T2	Α	Τ	Т
Т3	G	Т	G
T4	G	Т	G

discrete characters can be molecular or morphological

Inferring Phylogenies

we infer phylogenies by evaluating tree topologies

for 4 taxa we can evaluate all possible unrooted topologies (there are only 3)



How many trees?

n	Unrooted trees (U _n)	
3	1	
4	3	
5	15	
6	105	
7	945	
8	10, 395	
9	135, 135	
10	2,027,025	
20	~2.22 × 10 ²⁰	
50	~2.84 × 10 ⁷⁴	

$$N = \frac{(2t-5)!}{2^{t-3}(t-3)!}$$

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How many trees?

n	Unrooted trees (U _n)	Rooted trees (R _n)
3	1	3
4	3	15
5	15	105
6	105	945
7	945	10, 395
8	10, 395	135, 135
9	135, 135	2,027,025
10	2,027,025	34, 459, 425
20	~2.22 × 10 ²⁰	~8.20 × 10 ²¹
50	~2.84 × 10 ⁷⁴	~2.75 × 10 ⁷⁶

$$N = \frac{(2t-3)!}{2^{t-2}(t-2)!}$$

at 51 taxa, the number of trees exceeds the number of particles in the observable universe

How to find the "best" tree?



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It depends on how you measure "best"

Table 3.2 Optimality criteria used for phylogeny reconstruction

Method	Criterion (tree score)
Maximum parsimony	Minimum number of changes, minimized over ancestral states
Maximum likelihood	Log likelihood score, optimized over branch lengths and model parameters
Minimum evolution	Tree length (sum of branch lengths, often estimated by least squares)
Bayesian	Posterior probability, calculated by integrating over branch lengths and substitution parameters



Maximum Parsimony

the optimal tree is the one that has the fewest number of changes, given an observed set of discrete characters

based on the *parsimony principle*: assumes simpler explanations are better than complex ones



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Maximum Parsimony

to find the tree with the fewest changes:

- 1. construct all possible trees
- 2. count the minimum number of changes for every character in the matrix
- Sum the counts across all characters to obtain the "total tree length"
- 4. choose the tree with the lowest total tree length









Maximum Parsimony

computing the total tree length (parsimony score) is easy to do given a tree topology and observed character states (see Yang, 2014, ch. 3)

finding the tree topology that has the optimal parsimony score is hard (it is actually <u>NP-hard</u>)

Searching Tree Topologies

a number of heuristic tree-search algorithms have been developed (Yang 2014, Ch. 3)

these methods allow us to evaluate just a subset of the possible trees

Searching Tree Topologies

importantly, we need tree-search methods that can find the global optimum

these approaches are useful for maximum parsimony, maximum likelihood, and Bayesian methods

Parsimony and Assumptions

the parsimony principle is based on Occam's Razor: the simplest explanation that fits the data is preferred and *ad hoc* explanations should be avoided

parsimony does not make *explicit* assumptions about the evolutionary process that generated observed character states

Parsimony and Assumptions

with datasets comprising multiple characters there is typically no single topology that is the most parsimonious for every observation

thus, *ad hoc* explanations (convergence, reversals) must be invoked to explain these patterns

Homoplasy

A trait that is found in two species, but not in their common ancestor is an example of **homoplasy**

The eye structures of a human & a giant Pacific octopus are similar but evolved independently





Homoplasy and Parsimony



when characters conflict, ad hoc explanations (e.g., convergence) cannot be avoided

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Arguments Against Parsimony

parsimony does make *implicit* assumptions about evolutionary processes, though is is difficult to identify exactly what these are

parsimony has been demonstrated to be statistically inconsistent

for more on this, see Yang (2014), Ch. 5

Statistical Consistency

an estimator is consistent if it is guaranteed to get the correct answer with an infinite amount of data

we would prefer our estimators to be consistent



Parsimony can be Inconsistent

Felsenstein (1978) demonstrated that for some situations, parsimony is inconsistent and yields the wrong tree, even with an infinite amount of data

this issue is also known as **long-branch attraction** and is one of the strongest criticisms of the parsimony method

the probability of a parsimony informative site due to inheritance is very low



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(based on slides by <u>M.T. Holder</u>)

the probability of a misleading parsimony informative site due to parallelism is much higher



parsimony is almost guaranteed to get this tree wrong

parsimony will incorrectly place two long branches as sister lineages



under parsimony more change will be attributed to the internal branch

in the case of long branch attraction, parsimony is positively misleading





(based on slides by <u>M.T. Holder</u>)

Parsimony can be Inconsistent



the branch lengths (p, q) represent the probability of change along a branch



The Felsenstein Zone



the branch lengths (p, q) represent the probability of change along a branch





(Swofford et al. 2001)

Parsimony can be Inconsistent

if one feels that consistency is a desirable property for an estimator...

the inconsistency of parsimony is the strongest argument against its use

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