

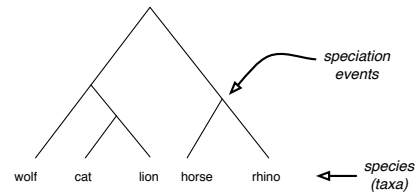
## Algorithms for Computational Biology

Zsuzsanna Lipták

Masters in Molecular and Medical Biotechnology  
a.a. 2015/16, fall term

### Phylogenetics Summary

## What is a phylogenetic tree?



Phylogenetic trees display the evolutionary relationships among a set of objects (species). Contemporary species are represented by the leaves. Internal nodes of the tree represent speciation events ( $\approx$  common ancestors, usually extinct).

2 / 45

## Different types of phylogenetic trees

- rooted vs. unrooted (root on top/bottom vs. root in the middle)
- binary (fully resolved) vs. multifurcating (polytomies)
- are edge lengths significant?
- is there a time scale on the side?

3 / 45

## Phylogenetic reconstruction

### Goal

Given  $n$  objects and data on these objects, find a phylogenetic tree with these objects at the leaves which best reflects the input data.

4 / 45

## Phylogenetic reconstruction

### Note:

We need to define more precisely

- what kind of input data we have,
- what kind of tree we want (e.g. rooted or unrooted), and
- what we mean by "reflect the data."

5 / 45

## Phylogenetic reconstruction

There are two main issues:

1. How well does a tree reflect my data?
2. How do we find such a tree?

6 / 45

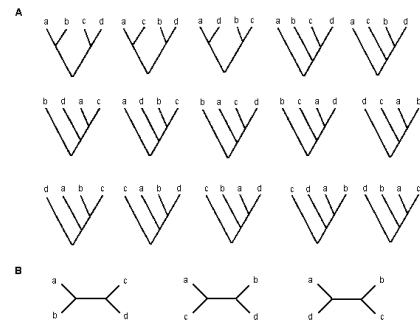
## Number of phylogenetic trees

Say we have answered these questions, then: Could we just list all possible trees and then choose the/a best one?

# taxa $n$	# unrooted trees $(2n - 5)!!$	# rooted trees $(2n - 3)!!$
1	1	1
2	1	1
3	1	3
4	3	15

7 / 45

## Number of phylogenetic trees



All phylogenetic trees (rooted and unrooted) on 4 taxa.

8 / 45

## Number of phylogenetic trees

### Theorem

There are  $U_n = (2n - 5)!! = \prod_{i=3}^n (2i - 5)$  unrooted binary phylogenetic trees on  $n$  objects, and  $R_n = (2n - 3)!! = \prod_{i=2}^n (2i - 3)$  rooted binary phylogenetic trees on  $n$  objects.

### Proof

By induction on  $n$ , using that (1) we can get every unrooted tree on  $n + 1$  objects in a unique way by adding the  $(n + 1)$ st leaf to an unrooted tree on the first  $n$  objects; (2) an unrooted binary tree with  $n$  leaves has  $2n - 3$  edges, (3) every unrooted tree on  $n$  objects can be rooted in (number of edges) ways, yielding a rooted tree on  $n$  objects.

9 / 45

## Number of phylogenetic trees

#taxa $n$	#unrooted trees $(2n - 5)!!$	#rooted trees $(2n - 3)!!$
1	1	1
2	1	1
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

10 / 45

## Number of phylogenetic trees

So there are **super-exponentially** many trees:  
We cannot check all of them!

11 / 45

## Types of input data

We can have two kinds of input data:

- **distance data:**  $n \times n$  matrix of pairwise distances between the taxa, or
- **character data:**  $n \times m$  matrix giving the states of  $m$  characters for the  $n$  taxa

12 / 45

## Distance data

**Distance data** is given as an  $(n \times n)$  matrix  $M$  with the pairwise distances between the taxa.

Ex.

	a	b	c
a	0	5	2
b	5	0	4
c	2	4	0

E.g.,  $M_{a,b} = 5$  means that the distance between  $a$  and  $b$  is 5. Often, this is the **edit distance** (between two genomic sequences, or between homologous proteins, ...).

We want to find a tree with  $a, b, c$  at the leaves s.t. the distance in the tree (the **path metric**) between  $a$  and  $b$  is 5, between  $a$  and  $c$  is 2, etc.

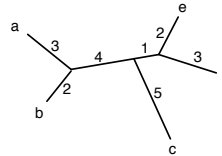
13 / 45

## Distance data

### Path metric of a tree

Given a tree  $T$ , the **path-metric** of  $T$  is  $d_T$ , defined as:  $d_T(u, v) =$  sum of edge weights on the (unique) path between  $u$  and  $v$ .

### Example



$$\begin{aligned} d_T(a, b) &= 5, \\ d_T(a, d) &= 11, \\ d_T(c, d) &= 9, \dots \end{aligned}$$

### Note

$d(u, v)$  is also defined for inner nodes  $u, v$ , but we only need it for leaves.

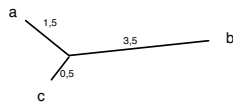
14 / 45

## Example

For our earlier example, we can find such a tree:

### Ex. 1 (from before)

	a	b	c
a	0	5	2
b	5	0	4
c	2	4	0



### Question

Is it always possible to find a tree s.t. its path-metric equals the input distances? I.e. does such a tree exist for **any** input matrix  $M$ ?

15 / 45

## Distance data

First of all, the input matrix  $M$  has to define a **metric** (= a distance function), i.e. for all  $x, y, z$ ,

- $M(x, y) \geq 0$  and  $(M(x, y) = 0 \text{ iff } x = y)$  (positive definite)
- $M(x, y) = M(y, x)$  (symmetry)
- $M(x, y) + M(y, z) \geq M(x, z)$  (triangle inequality)

For example, the **edit distance** is a metric (on strings), the **Hamming distance** (on strings of the same length), the **Euclidean distance** (on  $\mathbb{R}^2$ ).

16 / 45

## Conditions on distance matrix

### Question:

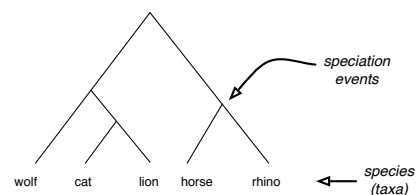
When does a tree exist whose path metric agrees with a distance matrix  $M$ ?

### Answer:

- if we want a **rooted** tree:  $M$  needs to be **ultrametric**
- if we want an **unrooted** tree:  $M$  needs to be **additive**

17 / 45

## Rooted trees and the molecular clock



In a rooted phylogenetic tree, the **molecular clock** assumption holds: that the speed of evolution is the same along all branches, i.e. the path distance from each leaf to the root is the same. Such a tree is also called an **ultrametric tree**.

18 / 45

## Ultrmetrics and the three-point condition

### Three point condition

Let  $d$  be a metric on a set of objects  $O$ , then  $d$  is an **ultrametric** if  $\forall x, y, z \in O$ :

$$d(x, y) \leq \max\{d(x, z), d(z, y)\}$$

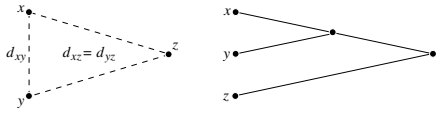


Figure : Three point condition. It implies that the path metric of a rooted tree is an ultrametric.

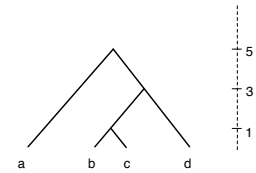
In other words, among the three distances, there is no unique maximum.

19 / 45

## Example

### Ex. 2

	a	b	c	d
a	0	10	10	10
b	10	0	2	6
c	10	2	0	6
d	10	6	6	0



Checking the ultrametric condition, we see that:

- for  $a, b, c$  we get 2, 10, 10 — okay
- for  $a, b, d$  we get 6, 10, 10 — okay
- for  $a, c, d$  we get 6, 10, 10 — okay
- for  $b, c, d$  we get 2, 6, 6 — okay

20 / 45

## Example

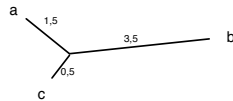
Compare this to our earlier example. There the matrix  $M$  does not define an ultrametric!

### Ex. 1 (from before)

	a	b	c
a	0	5	2
b	5	0	4
c	2	4	0

For the triple  $a, b, c$  (the only triple), we get: 2, 4, 5, and there is a unique maximum: 5.

Indeed, the only tree we found was not rooted:



21 / 45

## Ultrmetrics and the three-point condition

### Theorem

Given an  $(n \times n)$  distance matrix  $M$ . There is a rooted tree whose path metric agrees with  $M$  if and only if  $M$  defines an ultrametric (i.e. if and only if the 3-point-condition holds). This tree is unique<sup>1</sup>.

### Algorithm

The algorithm **UPGMA** (*unweighted pair group method using arithmetic averages*, Michener & Sokal 1957), a hierarchical clustering algorithm, constructs such a tree, given an input matrix which is ultrametric. Its running time is  $O(n^2)$ .

<sup>1</sup>i.e. there is only one such tree

22 / 45

## Additive metrics and the four-point condition

So what is the condition on the matrix  $M$  for unrooted trees?

### Four point condition.

Let  $d$  be a metric on a set of objects  $O$ , then  $d$  is an **additive metric** if  $\forall x, y, u, v \in O$ :

$$d(x, y) + d(u, v) \leq \max\{d(x, u) + d(y, v), d(x, v) + d(y, u)\}$$

In other words, among the three sums of two distances, there is no unique maximum.

23 / 45

## Additive metrics and the four-point condition

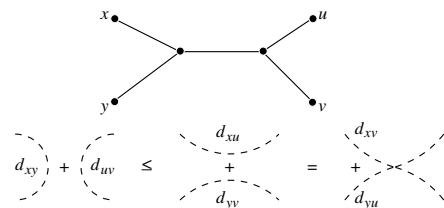
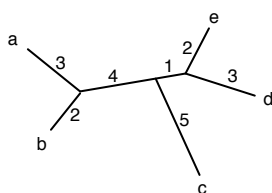


Figure : The four point condition. It implies that the path metric of a tree is an additive metric.

24 / 45

## Example



For ex., choose these 4 points:  $a, b, c, e$ . Then we get the three sums:  $d(a, b) + d(c, e) = 5 + 8 = 13$ ,  $d(a, c) + d(b, e) = 12 + 9 = 21$ , and  $d(a, e) + d(b, c) = 10 + 11 = 21$ . Among 13, 21, 21, there is no unique maximum—okay. (Careful, this has to hold for **all** quadruples; how many are there?)

25 / 45

## Additive metrics and the four-point condition

### Theorem

Given an  $(n \times n)$  distance matrix  $M$ . There is an unrooted tree whose path metric agrees with  $M$  if and only if  $M$  defines an additive metric (i.e. if and only if the 4-point-condition holds). This tree is unique.

### Algorithm

There are algorithms which, given  $M$ , compute this unrooted tree in  $O(n^3)$  time (e.g. Neighbor Joining, Saitu & Nei, 1987).

26 / 45

## Additive metrics and the four-point condition

### Theorem

Given an  $(n \times n)$  distance matrix  $M$ . There is an unrooted tree whose path metric agrees with  $M$  if and only if  $M$  defines an additive metric (i.e. if and only if the 4-point-condition holds). This tree is unique.

### Algorithm

There are algorithms which, given  $M$ , compute this unrooted tree in  $O(n^3)$  time (e.g. Neighbor Joining, Saitu & Nei, 1987).

In fact, it is even possible to compute a “good” tree if the matrix is not additive but “almost” (*all this needs to be defined precisely, of course*).

26 / 45

## Summary for distance data

- When the input is a **distance matrix**, then we are looking for a tree whose path metric agrees with  $M$ .
- A rooted tree agreeing with  $M$  exists if and only if the distance matrix  $M$  defines an ultrametric.
- This tree can then be computed efficiently (i.e. in polynomial time), with UPGMA - **which we studied!**
- An unrooted tree agreeing with  $M$  exists if and only if the distance matrix  $M$  defines an additive metric. It can then be computed efficiently (i.e. in polynomial time), with Neighbor Joining - **which we did not study.**

27 / 45

## Character data

Now the input data consists of **states of characters** for the given objects, e.g.

- morphological data, e.g. number of toes, reproductive method, type of hip bone, ... or
- molecular data, e.g. what is the nucleotide in a certain position.

28 / 45

## Character data

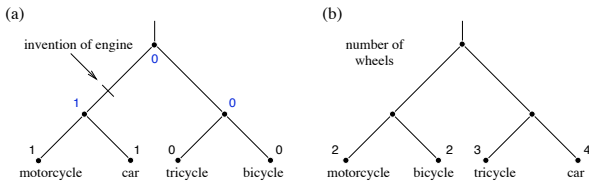
### Example

	$C_1$ : # wheels	$C_2$ : existence of engine
bicycle	2	0
motorcycle	2	1
car	4	1
tricycle	3	0

- **objects (species)**: Bicycle, motorcycle, tricycle, car
- **characters**: number of wheels; existence of an engine
- **character states**: 2, 3, 4 for  $C_1$ ; 0, 1 for  $C_2$  (1 = YES, 0 = NO)
- This matrix  $M$  is called a **character-state-matrix**, of dimension  $(n \times m)$ , where for  $1 \leq i \leq n, 1 \leq j \leq m$ :  $M_{ij}$  = state of character  $j$  for object  $i$ . (Here:  $n = 4, m = 2$ .)

29 / 45

## Character data



Two different phylogenetic trees for the same set of objects.

30 / 45

## Character data

We want to avoid

- parallel evolution (= convergence)
- reversals

Together these two conditions are also called **homoplasies**.

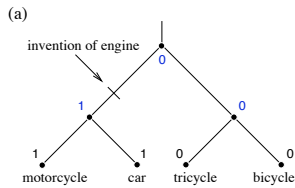
Mathematical formulation: **compatibility**.

31 / 45

## Compatibility

### Definition

A character is **compatible** with a tree if all inner nodes of the tree can be labeled such that each character state induces one connected subtree.



This tree is compatible with  $C_2$ , one possibility of labeling the inner nodes is shown.

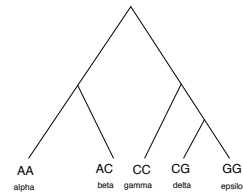
32 / 45

## Perfect Phylogeny

### Definition

A tree  $T$  is called a **perfect phylogeny (PP)** for  $\mathcal{C}$  if all characters  $C \in \mathcal{C}$  are compatible with  $T$ .

### Example



Why? We have to find a labeling of the inner nodes s.t. for both characters  $C_1$  and  $C_2$ , each state induces a subtree.

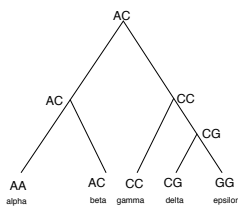
33 / 45

## Perfect Phylogeny

### Definition

A tree  $T$  is called a **perfect phylogeny (PP)** if all characters are compatible with  $T$ .

### Example



Our first tree for the vehicles was also a PP.

34 / 45

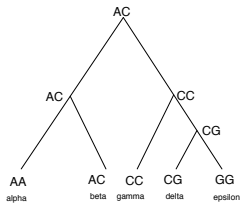
## Perfect Phylogeny

- Ideally, we would like to find a PP for our input data.
- Deciding in general whether a PP exists is NP-hard.
- This is not really a problem, since most of the time, no PP exists anyway. Why: due to homoplasies; because our input data has errors; our evolutionary model may have errors; and, and, and ...
- Therefore we usually want to find a **best possible** tree.

35 / 45

## Parsimony

What is a **best possible** tree?

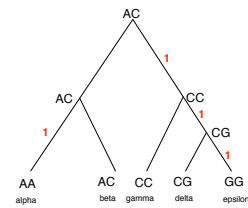


Why is this tree "perfect"?

36 / 45

## Parsimony

What is a **best possible** tree?



Why is this tree "perfect"?

Because it has few changes of states!

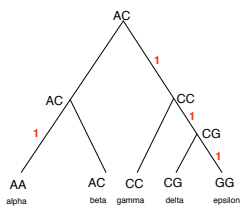
In red, we marked the edges where there are state changes (an evolutionary event happened), and how many (in this case, always 1).

37 / 45

## Parsimony

### Definition

The **parsimony cost** of a phylogenetic tree **with labeled inner nodes** is the number of state changes along the edges (i.e. the sum of the edge costs, where the cost of an edge = number of characters whose state differs between child and parent).



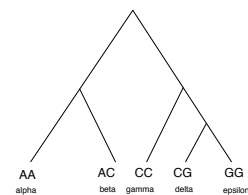
The parsimony cost of this labeled tree is 4.

38 / 45

## Parsimony

### Definition

The **parsimony cost** of a phylogenetic tree (**without labels on the inner nodes**) is the **minimum** of the parsimony cost over all possible labelings of the inner nodes.



The parsimony cost of this tree is 4, because the best labeling has cost 4.

39 / 45

## Parsimony

### Phylogenetic Reconstruction with Character Data

Given a character-state matrix  $M$ , our goal is to find a phylogenetic tree which minimizes the parsimony cost.

Given the character-state matrix  $M$ , we split the problem in two sub-problems:

1. **Small Parsimony:** Given a phylogenetic tree, find its parsimony cost, i.e. find a most parsimonious labeling of the inner nodes. This problem can be solved efficiently.
2. **Large Parsimony** or **Maximum Parsimony:** Find a tree with minimum parsimony cost. This problem is NP-hard.

40 / 45

## Small Parsimony

### Small Parsimony Problem

**Given:** a phylogenetic tree  $T$  with character-states at the nodes.

**Find:** a labeling of the inner nodes with states with minimum parsimony cost.

### Algorithm

This problem can be solved using **Fitch' algorithm**, which runs in time  $O(nmr)$ , where  $n$  = number of species,  $m$  = number of characters, and  $r$  = maximum number of states over all characters.

41 / 45

## Maximum Parsimony

### Maximum Parsimony Problem

The **maximum parsimony problem** is, given a character-state matrix, find a phylogenetic tree with lowest parsimony cost (= a "**most parsimonious tree**").

- When a PP exists, then it is also the most parsimonious tree.
- In general, this problem is NP-hard.

42 / 45

## Algorithms for Maximum Parsimony

- Since problem NP-hard, we cannot hope to find an algorithm that solves it efficiently.
- We have seen two algorithms for this problem:
  1. **Greedy Sequential Addition Algorithm** - heuristic algorithm: guaranteed polynomial running time but no guarantee on the quality of the solution
  2. **Branch-and-Bound for Parsimony** - runtime heuristic: guarantee on exact solution, but no guarantee on the running time (may or may not be fast)

43 / 45

## Summary for character data

- When the input is a **character-state matrix**, then we would like to find a tree which is compatible with each character.
- Such a tree is called a **perfect phylogeny (PP)**.
- Usually, no PP exists, therefore in general ...
- We are looking for a **most parsimonious tree** (a tree with lowest parsimony cost).
- The parsimony cost is defined as the minimum number of the state changes on the edges **over all possible labelings of the inner nodes**.
- **Recall:** There are super-exponentially many trees on  $n$  taxa (both rooted and unrooted), so we cannot try them all.

44 / 45

## Summary for character data (cont'ed)

- **recall:** We are looking for a **most parsimonious tree** (a tree with lowest parsimony cost).
- Problem is split into **Small Parsimony** and **Maximum Parsimony**.
- **Small Parsimony** can be solved efficiently, e.g. by Fitch' algorithm.
- **Maximum Parsimony** is NP-hard, so probably no efficient algorithms exist.
- We saw two algorithms for Maximum Parsimony: one heuristic (Greedy Seq. Add. Alg.) and one exact algorithm which is a runtime heuristic algorithm (Branch-and-Bound for Parsimony).

45 / 45