Bioinformatics Algorithms

(Fundamental Algorithms, module 2)

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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 nucletoide in a certain position.

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* × *m*), where for 1 ≤ *i* ≤ *n*, 1 ≤ *j* ≤ *m*: *M*_{ij} = state of character *j* for object *i*. (Here: *n* = 4, *m* = 2.)



Two different phylogenetic trees for the same set of objects.

We want to avoid

- parallel evolution (= convergence)
- reversals

Together these two conditions are also called homoplasies.

Mathematical formulation: compatibility.

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.

Definition

A tree T is called a perfect phylogeny (PP) for C if all characters $C \in C$ 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.

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Note: Our first tree for the vehicles was also a PP. (Proof?)

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- Therefore we usually want to find a best possible tree.

Compatibility and Parsimony

What is a **best possible** tree?

Two possibilities:

- Compatibility: what is the largest subset of the characters such that a PP exists? This means ignoring part of the input data.
- Parsimony: If we want to keep our input data, then what is the smallest number of changes that have to be made along the edges?

Parsimony: What is a best possible tree?



Why is this tree "perfect"?

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).

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.

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.

Phylogenetic Reconstruction with Character Data

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

We split the problem into 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.

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.

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.

Algorithms for Maximum Parsimony

- Since problem NP-hard, we cannot hope to find an algorithm that solves it efficiently.
- We have seen the following algorithms for this problem:
 - 1. Greedy Sequential Addition Algorithm heuristic algorithm: guaranteed polynomial running time but no guarantee on the quality of the solution (may or may not be correct, i.e. may or may not output the best tree)
 - Branch-and-Bound for Parsimony running time heuristic: guarantee on exact solution, but no guarantee on the running time (may or may not be fast)
 - 3. Nearest Neighbor Interchange a local optimization algorithm (also a heuristic algorithm, but guarantees to output a local optimum)

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- Recall: There are super-exponentially many trees on *n* taxa (both rooted and unrooted), so we cannot try them all.

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- Problem is split into Small Parsimony and Maximum Parsimony.
- Small Parsimony can be solved efficienly, e.g. by Fitch' algorithm.
- Maximum Parsimony is NP-hard, so probably no efficient algorithms exist.
- We saw three algorithms for Maximum Parsimony: one heuristic (Greedy Seq. Addition Algo.) and one exact algorithm which is a running time heuristic (Branch-and-Bound for Parsimony), and a local optimization algorithm (Nearest Neighbor Interchange).