Which Animal Gave Us SARS – Part 2

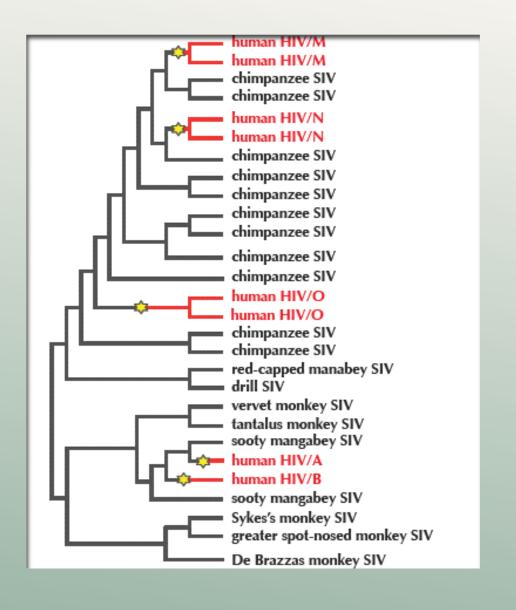
Evolutionary tree Reconstruction



Selected Topics in Bioinformatics Lecture 4

AGENDA

Computing limb lengths Trimming the tree Attaching a limb



Definitions revision

- In a tree, nodes with degree larger than 1 are called **internal nodes** while, those with degree 1 are called a **leaves**.
- Fiven a leaf j, there is only one node connected to j by an edge called the **parent** of j, denoted **PARENT(j)**.
- An edge connecting a leaf to its parent is called a <u>limb</u>.
- A **rooted tree** is a tree with one node designated as a special node called the **root.**

Another approach for Neighboring leaves

- The assumption that a minimum element of an additive distance matrix corresponds to neighboring leaves is not necessarily true!
- Thus, we need a new approach to the Distance-Based Phylogeny Problem
- > so, we will explore a different recursive algorithm.
 - Rather than looking for a pair of neighbors in TREE (D),
 - ► Instead, we will reduce the size of the tree by trimming its leaves one at a time.
 - Of course, we don't know TREE (D)
 - So, we must somehow trim leaves in TREE(D) by analyzing the distance matrix.

Another approach for Neighboring leaves

- The first step towards constructing TREE (D), is to compute the lengths of limbs in TREE (D)
- Given a leaf j in a tree, we denote the length of the limb connecting j with its parent as LIMBLENGHT (j).
- Edges that are not limbs must connect two internal nodes and are therefore called internal edges.

Another approach for Neighboring leaves

Limb Length Problem:

Compute the length of a limb in the simple tree fitting an additive distance matrix.

Input: An additive distance matrix D and an integer j.

Output: LIMBLENGTH(j), the length of the limb connecting leaf j to its

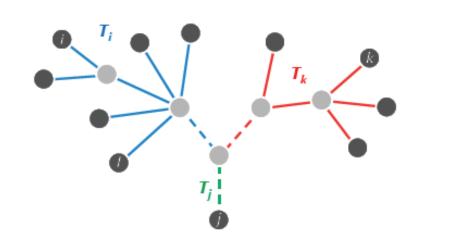
parent in TREE(D).

Limb Length Theorem

- When computing LIMBLENGHT (*j*) for a given leaf *j*, and TREE (D) is a simple tree, then PARENT (*j*) has degree at least 3.
- ▶ PARENT (*j*) as dividing the other nodes of TREE (D) into at least 3 subtrees,

(i.e. smaller trees that would remain if we remove PARENT (j) with any edges connecting it to other nodes)

 \triangleright Because *j* is a leaf, it must belong to a subtree by itself; we call it T_i .



Limb Length Theorem

Limb Length Theorem: Given an additive matrix D and a leaf j, LIMBLENGTH(j) is equal to the minimum value of $(D_{i,j} + D_{j,k} - D_{i,k})/2$ over all leaves i and k.

i.e., For each leave j, we can compute LIMBLENGHT (j) by finding the minimum value of $(D_{i,j} + D_{j,k} - D_{i,k})/2$ over all pairs of leaves i and k.

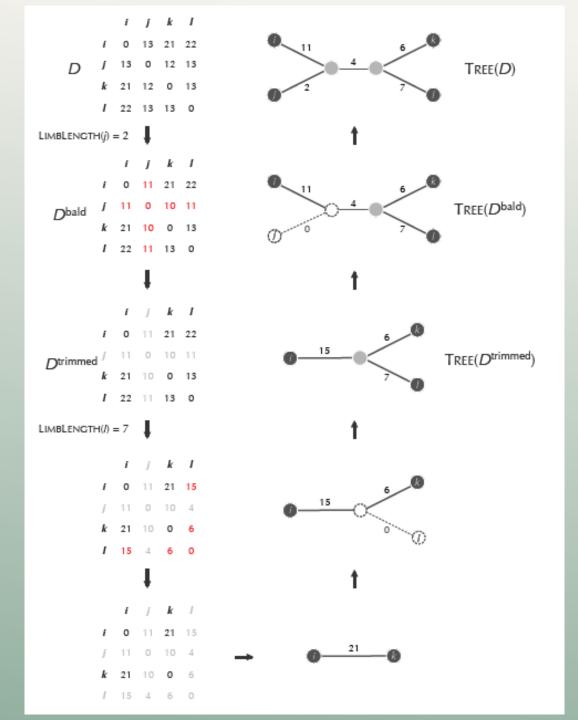
TRIMMING THE TREE

A New Algorithm for reconstructing the TREE

- Since we now know how to find the length of any limb in TREE(D), we can construct TREE (D) recursively using a new algorithm
- pick an arbitrary leaf j, compute LIMBLENGTH(j), and construct the distance matrix D^{trimmed};
- solve the Distance-Based Phylogeny Problem for D^{trimmed};
- identify the point in TREE(D^{trimmed}) where leaf j should be attached in TREE(D);
- add a limb of length LIMBLENGTH(j) growing from this attachment point in TREE(D^{trimmed}) to form TREE(D).

TRIMMING THE TREE

- First compute LIMBLENGHT (j) = 2,
- Subtract 2 from the non-diagonal elements in row j and column j of D to obtain D^{bald} (updated values are shown in red).
- Removing this row and column yields a 3 × 3 distance matrix D^{trimmed}.
- We find that LIMBLENGHT (l) = 7 in D^{trimmed}
- Subtract 7 from the non-diagonal elements in row l and column l.
- Graying out this row and column yields a 2 × 2 distance matrix.
- On the right side, we fit this 2×2 distance matrix to a tree consisting of a single edge.
- By finding the attachment points of removed limbs (shown on the left), we reconstruct TREE (D^{trimmed}), TREE (D^{bald}), and then TREE (D).



ATTACHING A LIMB

How to find the attachment point of leaf j in TREE (D^{trimmed})?

- TREE (D^{bald}) is the same as TREE (D) except that LIMBLENGHT (j) = 0.
- From the Limb Length Theorem, there must be leaves i and k in TREE (D^{bald}) such that:

$$\frac{D_{i,j}^{\text{bald}} + D_{j,k}^{\text{bald}} - D_{i,k}^{\text{bald}}}{2} = 0,$$

which implies that

$$D_{i,k}^{\text{bald}} = D_{i,j}^{\text{bald}} + D_{j,k}^{\text{bald}}.$$

- Thus, the attachment point for leaf j must be located at distance $D^{bald}_{i,j}$ from leaf i on the path connecting i and k in the trimmed tree.
- This attachment point may occur at an existing node, in which case we connect *j* to this node.
- ➤ On the other hand, the attachment point for *j* may occur along an edge, in which case we place a new node at the attachment point and connect *j* to it.

ATTACHING A LIMB

Distance-based phylogeny construction Algorithm

```
ADDITIVE PHYLOGENY (D, n)
    if n=2
        return the tree consisting of a single edge of length D_{1,2}
    limbLength \leftarrow Limb(D, n)
    for j \leftarrow 1 to n-1
        D_{j,n} \leftarrow D_{j,n} - limbLength
        D_{n,j} \leftarrow D_{j,n}
    (i, n, k) \leftarrow three leaves such that D_{i,k} = D_{i,n} + D_{n,k}
   x \leftarrow D_{i,n}
    remove row n and column n from D
    T \leftarrow AdditivePhylogeny(D, n-1)
    v \leftarrow the (potentially new) node in T at distance x from i on the path between i and k
    add leaf n back to T by creating a limb (v, n) of length limbLength
    return T
```

New Goal

- Given a non-additive $n \times n$ distance matrix D, then
 - instead of looking for a weighted tree T whose distances between leaves approximate the entries in D.
 - We want look for T that minimize the sum of squared errors DISCREPANCY (T, D), which is given by the formula

DISCREPANCY
$$(T, D) = \sum_{1 \le i < j \le n} (d_{i,j}(T) - D_{i,j})^2$$
.

USING LEAST SQUARE TO CONSTRUCT PHYLOGENY Least Squares Distance-Based Phylogeny Problem

Least Squares Distance-Based Phylogeny Problem:

Given a distance matrix, find the tree that minimizes the sum of squared errors.

Input: An $n \times n$ distance matrix D.

Output: A weighted tree T minimizing DISCREPANCY(T, D) over all weighted trees with n leaves.

Challenges for this problem

- For a specific tree T, it is easy to find edge weights in T minimizing DISCREPANCY (T, D).
- But minimizing the sum of squared errors for a specific tree does not imply that we can efficiently solve the Least Squares Distance-Based Phylogeny Problem, since the number of different trees grows very quickly as the number of leaves in the tree increases.
- In fact, the Least Squares Distance-Based Phylogeny is an NP-Complete problem
 - So, we look for a heuristics for constructing trees from non-additive matrices that solve this problem *approximately*.

Rooted/Unrooted Binary Trees

- Biologists assume that every internal node in an evolutionary tree corresponds to a species that underwent a speciation event:
 - > Splitting one ancestral species into two descendants.
- So, we define an **unrooted binary tree** as a tree where every node has degree equal to either 1 or 3.
- A **rooted binary tree** is an unrooted binary tree that has a root (of degree 2) placed on one of its edges.
 - we replace an edge (v, w) with a root and draw edges connecting the root to each of v and w

Ultrametric Evolutionary Trees

- We assign an **age** to every node v in a rooted binary tree (denoted AGE(v)), where all the leaves of the tree have age o
 - because they correspond to present-day species.
- We define the weight of an edge (v, w) in the tree as the difference AGE(v) AGE(w).
- The length of a path between the root and any node would be equal to the difference between their ages.
- A tree, in which the distance from the root to any leaf is the same, is called **ultrametric**.

Ultrametric Evolutionary Trees

- Our aim is to derive an ultrametric tree that explains a distance matrix (even if it is only approximately).
- **UPGMA** (Unweighted Pair Group Method with Arithmetic Mean) is a simple clustering heuristic construct an ultrametric evolutionary tree.

UPGMA

- Given an $n \times n$ matrix D, UPGMA first forms n trivial clusters, each containing a single leaf.
- ▶ The algorithm then finds a pair of "closest" clusters.
- To calculate closest clusters, UPGMA defines the distance between clusters *C1* and *C2* as the average pairwise distance between elements of *C1* and *C2*.

$$D_{C_1,C_2} = \frac{\sum_{i \in C_1} \sum_{j \in C_2} D_{i,j}}{|C_1| \cdot |C_2|}.$$

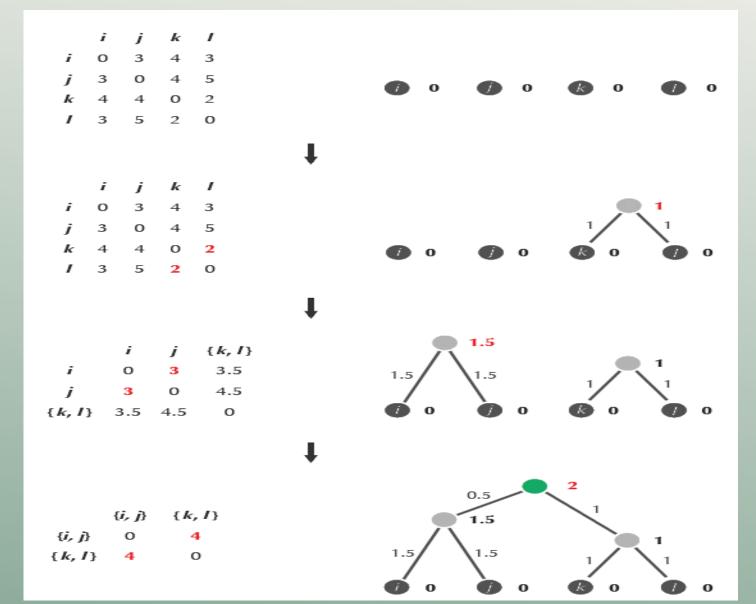
UPGMA

- |C| denotes the number of leaves in cluster C.
- Once a pair of closest clusters C1 and C2 are identified, merges them into a cluster C with |C1|+|C2| elements and then creates a node for C, which it connects to each of C1 and C2 by a directed edges.
- The age of C is set to be $D_{C_1,C_2}/2$.
- UPGMA then iterates this process of merging the two closest clusters until only a single cluster remains, which corresponds to the root.

UPGMA Algorithm

```
UPGMA(D, n)
    Clusters \leftarrow n single-element clusters labeled 1, \ldots, n
    construct a graph T with n isolated nodes labeled by single elements 1, \ldots, n
    for every node v in T
       AGE(v) \leftarrow 0
    while there is more than one cluster
       find the two closest clusters C_i and C_j (break ties arbitrarily)
       merge C_i and C_j into a new cluster C_{\text{new}} with |C_i| + |C_j| elements
       add a new node labeled by cluster C_{new} to T
       connect node C_{\text{new}} to C_i and C_i by directed edges
       AGE(C) \leftarrow D_{C_i,C_i}/2
       remove the rows and columns of D corresponding to C_i and C_i
       remove C_i and C_i from Clusters
       add a row/column to D for C_{\text{new}} by computing D(C_{\text{new}}, C) for each C in Clusters
       add Cnew to Clusters
   root \leftarrow the node in T corresponding to the remaining cluster
    for each edge (v, w) in T
       length of (v, w) \leftarrow AGE(v) - AGE(w)
    return T
```

Example:



Disadvantages of Using the Neighbors

- The first step that UPGMA is to merge the two leaves i and j with minimum distance $D_{i,j}$ into a single cluster.
- And we have already seen that the smallest element in the distance matrix does not necessarily correspond to a pair of neighboring leaves!
- This is a concern, since if UPGMA generates incorrect trees from additive matrices, then it is not an ideal heuristic for evolutionary tree construction from non-additive matrices.
- Can we find an algorithm that always identifies neighboring leaves in an additive distance matrix but also performs well on a non-additive distance matrix?