CMSC423: Bioinformatic Algorithms, Databases and Tools Lecture 12

Phylogenetic trees
Phylogenetic tree display
Phylogenetic analysis

## Sankoff's algorithm

- At each node $v$ in the tree store $s(v, t)$ - best parsimony score for subtree rooted at $v$ if character stored at v is t
- Traverse the tree in post-order and update $s(v, t)$ as follows
- assume node $v$ has children $u$ and w
$-\mathrm{s}(\mathrm{v}, \mathrm{t})=\min _{\mathrm{i}}\{\mathrm{s}(\mathrm{u}, \mathrm{i})+\mathrm{score}(\mathrm{i}, \mathrm{t})\}+\min _{\mathrm{j}}\{\mathrm{s}(\mathrm{w}, \mathrm{j})+\mathrm{score}(\mathrm{j}, \mathrm{t})\}$
- Character at root will be the one that maximizes s (root, t)
- Note - this solves the weighted version. For unweighted set score (i,i) = 0, score(i,j) = 1 for any $i, j$


## Sankoff's algorithm - example

$$
s(v, t)=\operatorname{mini}\{s(u, i)+\operatorname{score}(i, t)\}+\operatorname{minj}\{s(w, j)+\operatorname{score}(j, t)\}
$$



## Trees as clustering

- Start with a distance matrix - distance (e.g. alignment distance) between any two sequences (leaves)
- Intuitively - want to cluster together the most similar sequences
- UPGMA - Unweighted Pair Group Method using Arithmetic averages
- Build pairwise distance matrix (e.g. from a multiple alignment)
- Pick pair of sequences that are closest to each other and cluster them create internal node that has the sequences as children
- Repeat, including newly created internal nodes in the distance matrix

- Key element - must be able to quickly compute distance between clusters (internal nodes) - weighted distance

$$
D\left(c l_{1}, c l_{2}\right)=\frac{1}{\left|c l_{1}\right|\left|c l_{2}\right|} \sum_{p \in c l_{1}, q \in l_{2}} D(p, q)
$$

## Trees as clustering

- Note that UPGMA does not estimate branch lengths - they are all assumed equal
- Neighbor-joining
- distance between two sequences is not sufficient - must also know how each sequence compares to every other sequence
- NJdist $(\mathrm{i}, \mathrm{j})=\mathrm{D}(\mathrm{i}, \mathrm{j})-\left(\mathrm{r}_{\mathrm{i}}+\mathrm{r}_{\mathrm{j}}\right) \quad-\mathrm{r}_{\mathrm{i}}, \mathrm{r}_{\mathrm{j}}$ correction factors

$$
r_{i}=\frac{1}{m-2} \sum_{k} D(i, k)
$$

- Pick two nodes with NJdist(i,j) minimal
- Create parent k s.t.
- $D(k, m)=0.5(D(i, m)+D(j, m)-D(i, j))$ for every other node $m$
$-D(i, k)=0.5\left(D(i, j)+r_{i}-r_{j}\right)$ - length of branch between i \&
$-D(j, k)=0.5\left(D(i, j)+r_{j}-r_{i}\right)$ length of branch between $j \& k$


## Trees as clustering

- Note that both UPGMA and NJ assume distance matrix is additive: $\mathrm{D}(\mathrm{i}, \mathrm{j})+\mathrm{D}(\mathrm{j}, \mathrm{k})=\mathrm{D}(\mathrm{i}, \mathrm{k})$ - usually not true but close
- Also, NJ can be proven to build the optimal tree!
- But, simple alignment distance is not a good metric


## Maximum likelihood

- For every branch S->T of length $t$, compute $P(T \mid S, t)$ likelihood that sequence $S$ could have evolved in time t into sequence T
- Find tree that maximizes the likelihood
- Note that likelihood of a tree can be computed with an algorithm similar to Sankoffs
- However, no simple way to find a tree given the sequences - most approaches use heuristic search techniques
- Often, start with NJ tree - then "tweak" it to improve likelihood


## From multiple alignment to tree




## Three types of trees

Cladogram


Phylogram


All show the same evolutionary relationships, or branching orders, between the taxa.
from www.albany.edu/faculty/cs812/StewartTalk2.ppt

