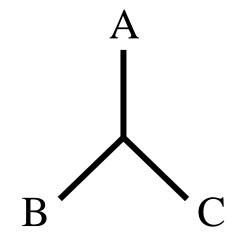
We've discussed how we rank trees

- Parsimony
- Least squares
- Minimum evolution
- Balanced minimum evolution
- Maximum likelihood (later in the course)

So we have ways of deciding what a good tree is when we see one, but . . .

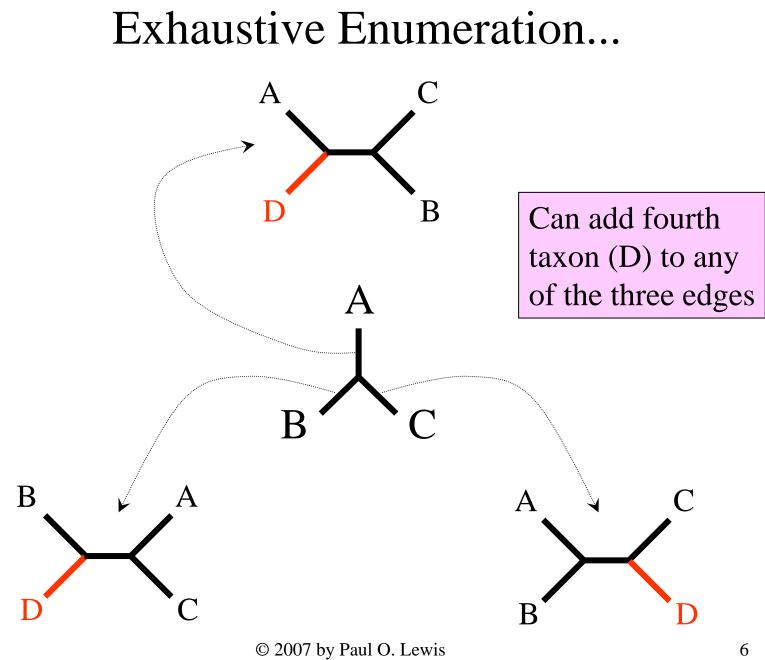
> How do we find the best tree? (or one that is good enough)

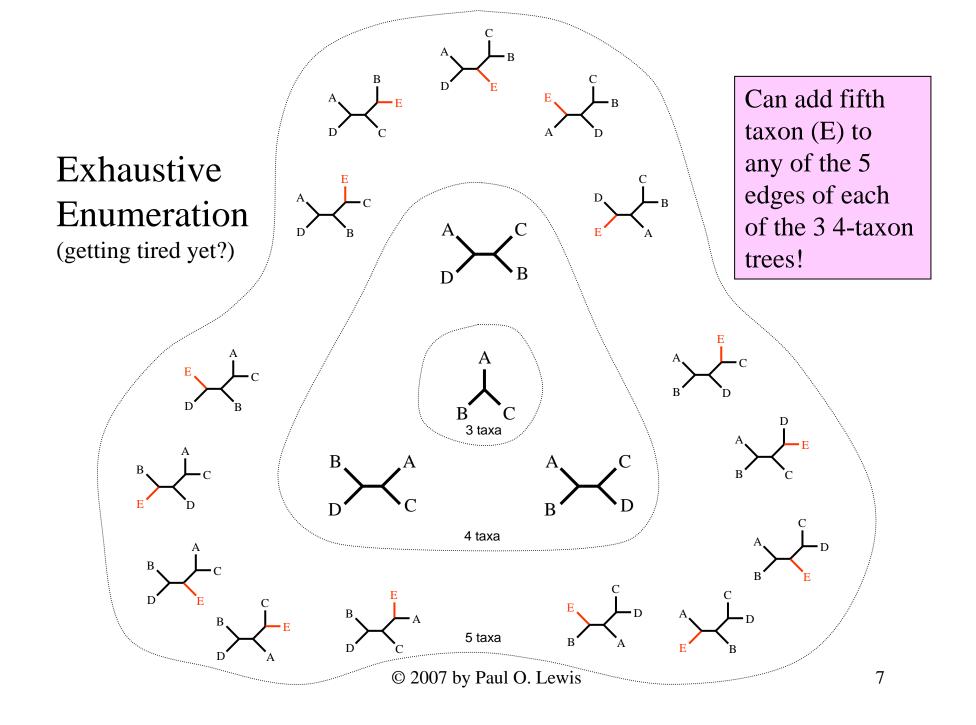
Exhaustive Enumeration



With the first three taxa, create the trivial unrooted tree

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Tips	Number of unrooted (binary) trees	
4	3	
5	15	
6	105	
7	945	
8	10,395	
9	135,135	
10	2,027,025	
11	34,459,425	
12	654,729,075	
13	13,749,310,575	
14	316,234,143,225	
15	7,905,853,580,625	
16	213,458,046,676,875	
17	6,190,283,353,629,375	
18	191,898,783,962,510,625	
19	6,332,659,870,762,850,625	
20	22,164,309,5476,699,771,875	
21	8,200,794,532,637,891,559,375	
22	319,830,986,772,877,770,815,625	
23	13,113,070,457,687,988,603,440,625	>21 moles of trees
24	563,862,029,680,583,509,947,946,875	

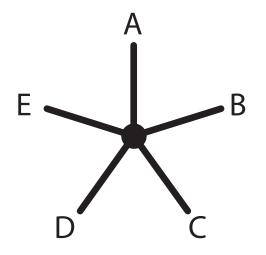
For N taxa:

unrooted, binary trees =
$$\prod_{i=3}^{N-1} (2i-3)$$

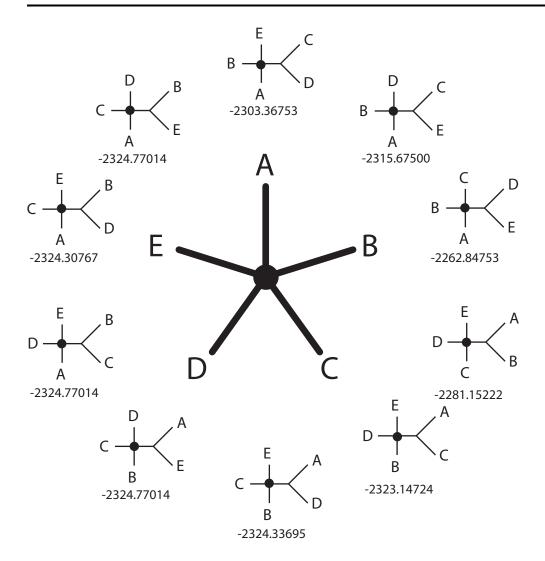
=
$$\prod_{i=4}^{N} (2i-5)$$

rooted, binary trees =
$$\prod_{i=3}^{N} (2i-3)$$

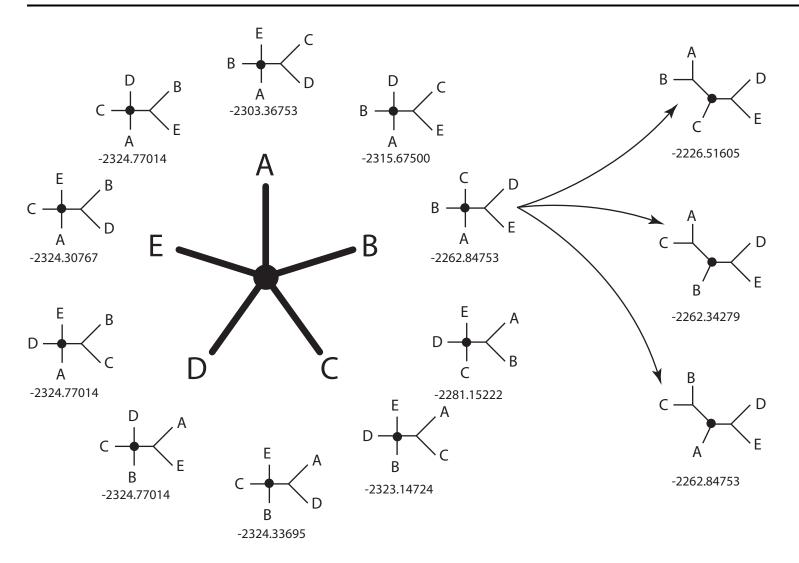
=
$$(2N-3)(\text{# unrooted, binary trees})$$



Star decomposition

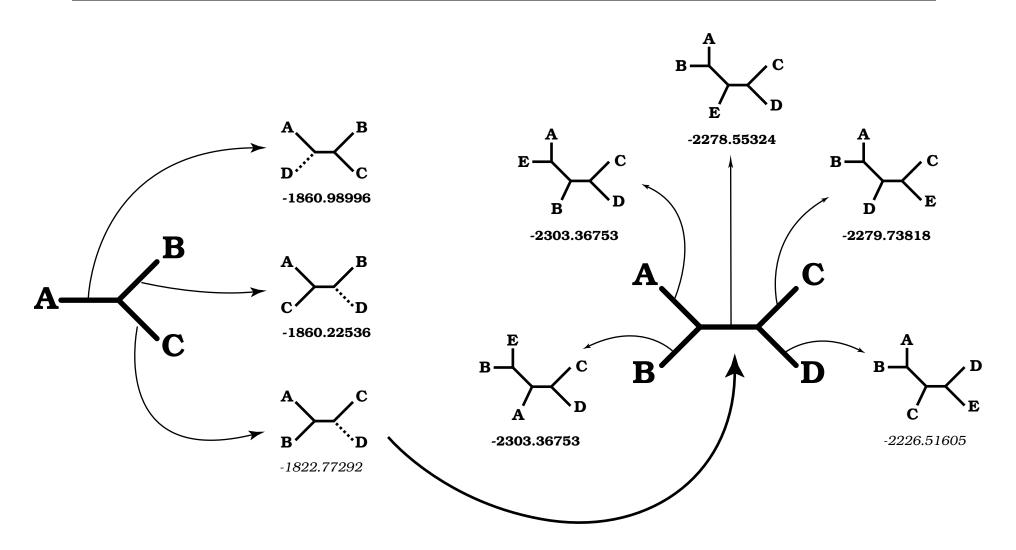


Star decomposition



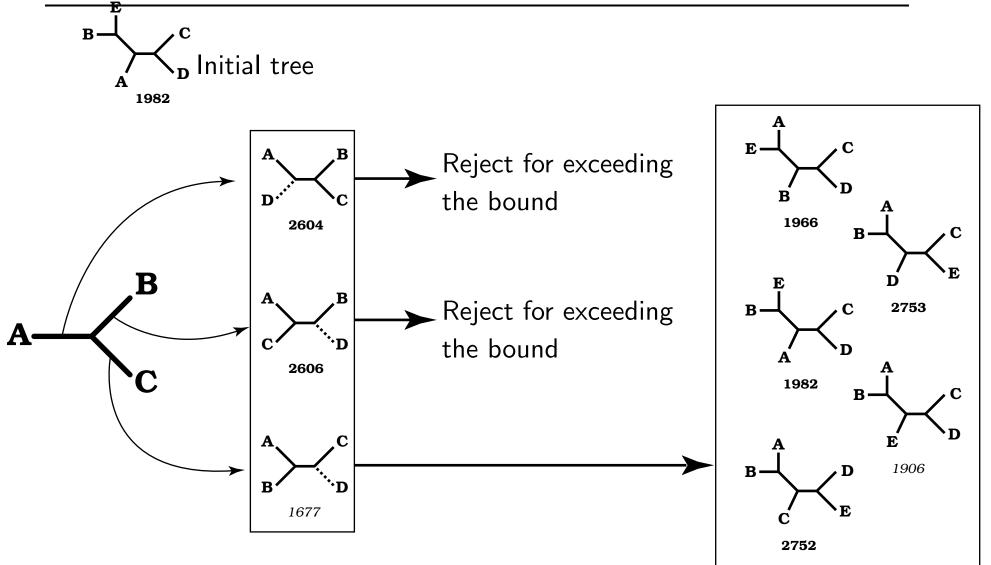
- Very "greedy" it makes the best decision at each step, but does not try to "plan ahead". Once a pair of species are joined, they will not be separated.
- Neighbor-joining (Saitou and Nei, 1987) is star decomposition under the balanced minimum evolution criterion

Stepwise addition



- Order-dependent (multiple random orderings can be used to give a range of starting trees for more thorough searches).
- Taxa joined initially may have intervening species added, but still fairly greedy.

Branch and bound



- Guaranteed to return the best tree(s)
- Typically only a viable option for < 30 species (depends on how clean the data is)

Neither stepwise addition nor star decomposition is guaranteed to return the best tree(s), but branch-and-bound (or exhaustive searching) is frequently infeasible.

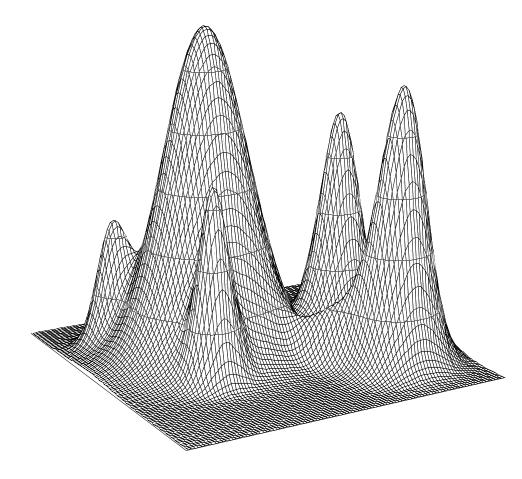
Heuristic hill-climbing searches can work quite well:

- 1. Start with a tree
- 2. Score the tree
- 3. Consider a new tree within the neighborhood of the current tree:
 - (a) Score the new tree.
 - (b) If the new tree has a better tree, use it as the "current tree"
 - (c) Stop if there are no other trees within the neighborhood to consider.

These are **not** guaranteed to find even one of the optimal trees.

The most common way to explore the neighborhood of a tree is to swap the branches of the tree to construct similar trees.

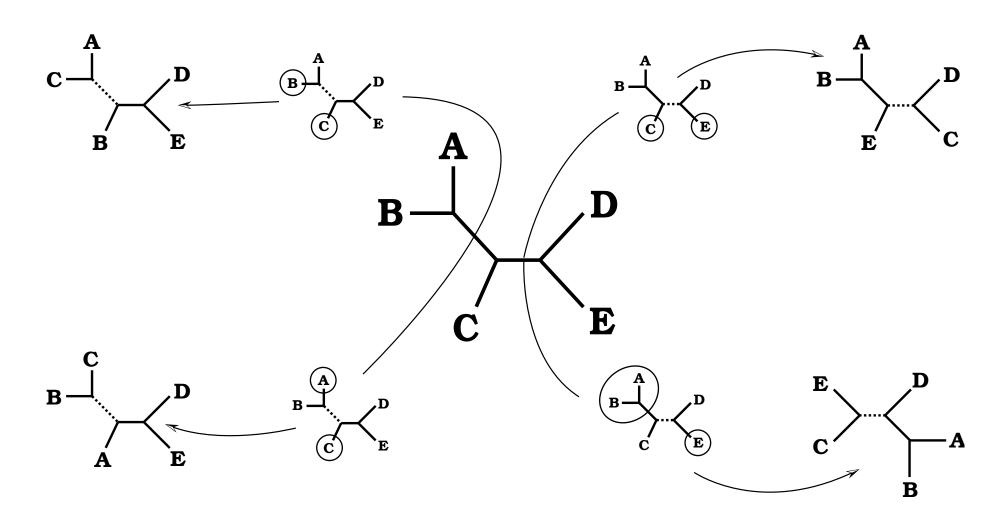
Heuristics explore "Tree Space"



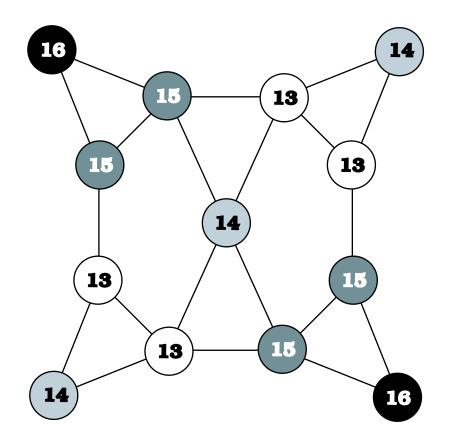
Most commonly used methods are "hill-climbers."

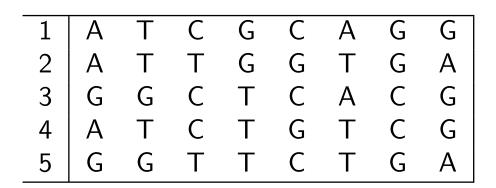
Multiple optima found by repeating searches from different origins.

Severity of the problem of multiple optima depends on step size.



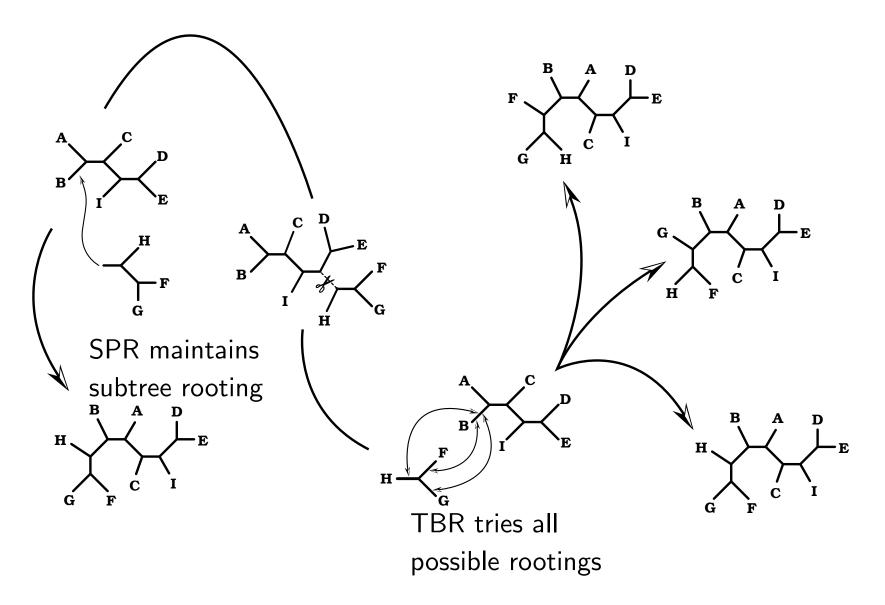
Nearest Neighbor Interchange (NNI)





Contrived matrix with 2 NNI islands

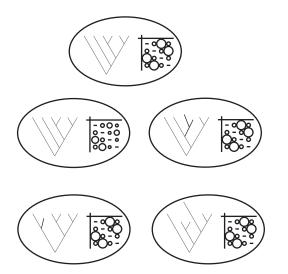
Subtree Pruning Regrafting (SPR) and Tree Bisection Reconnection (TBR)

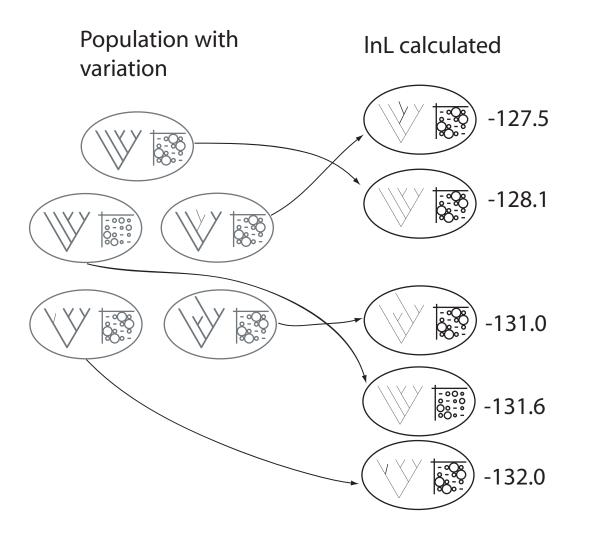


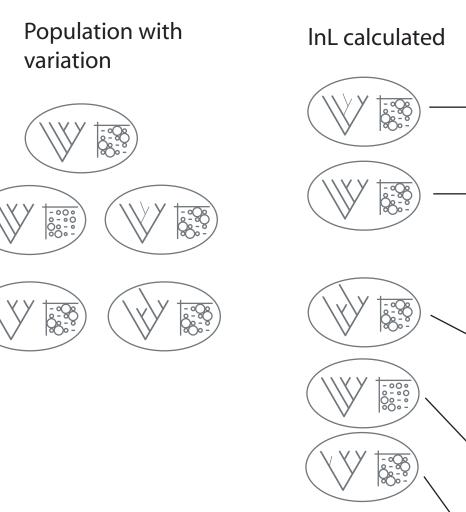
Many other heuristic strategies proposed

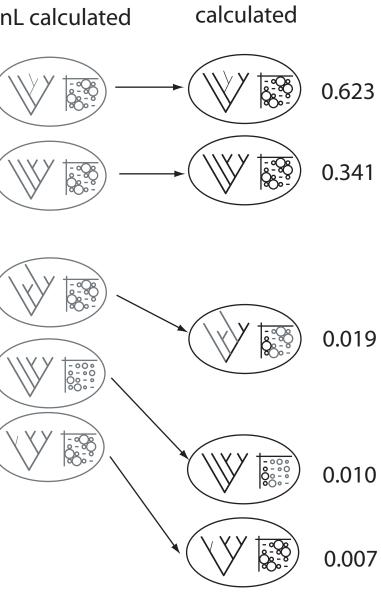
- Swapping need not include *all* neighbors (RAxML, reconlimit in PAUP*)
- "lazy" scoring of swaps (RAxML)
- Ignoring (at some stage) interactions between different branch swaps (PHYML)
- Stochastic searches
 - Genetic algorithms (GAML, MetaPIGA, GARLI)
 - Simulated annealing
- Divide and conquer methods (the sectortial searching of Goloboff, 1999; Rec-I-DCM3 Roshan 2004)
- Data perturbation methods (e.g. Kevin Nixon's "ratchet")

Population with variation

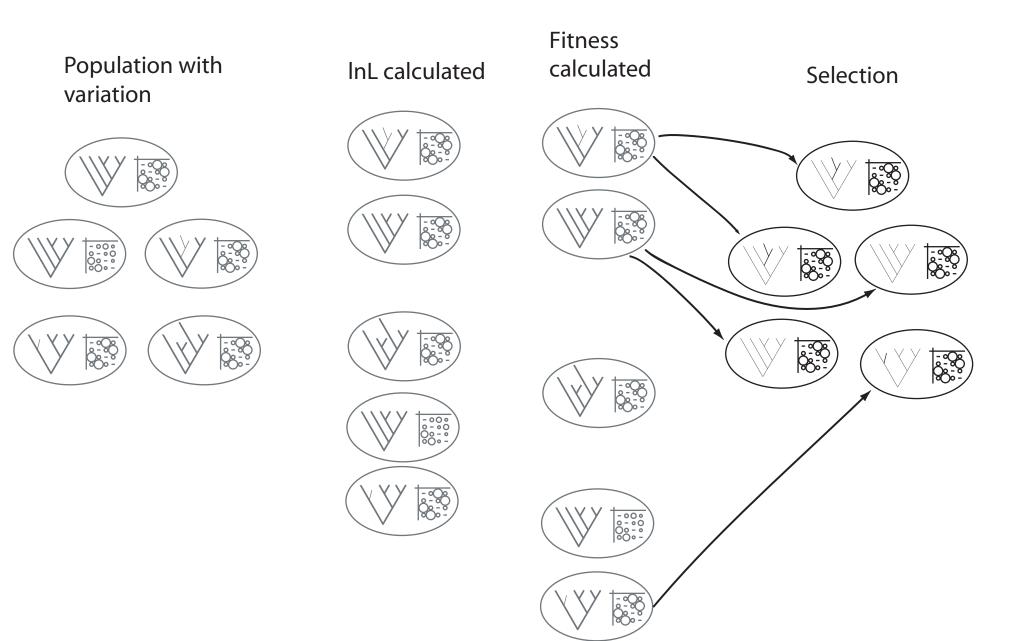


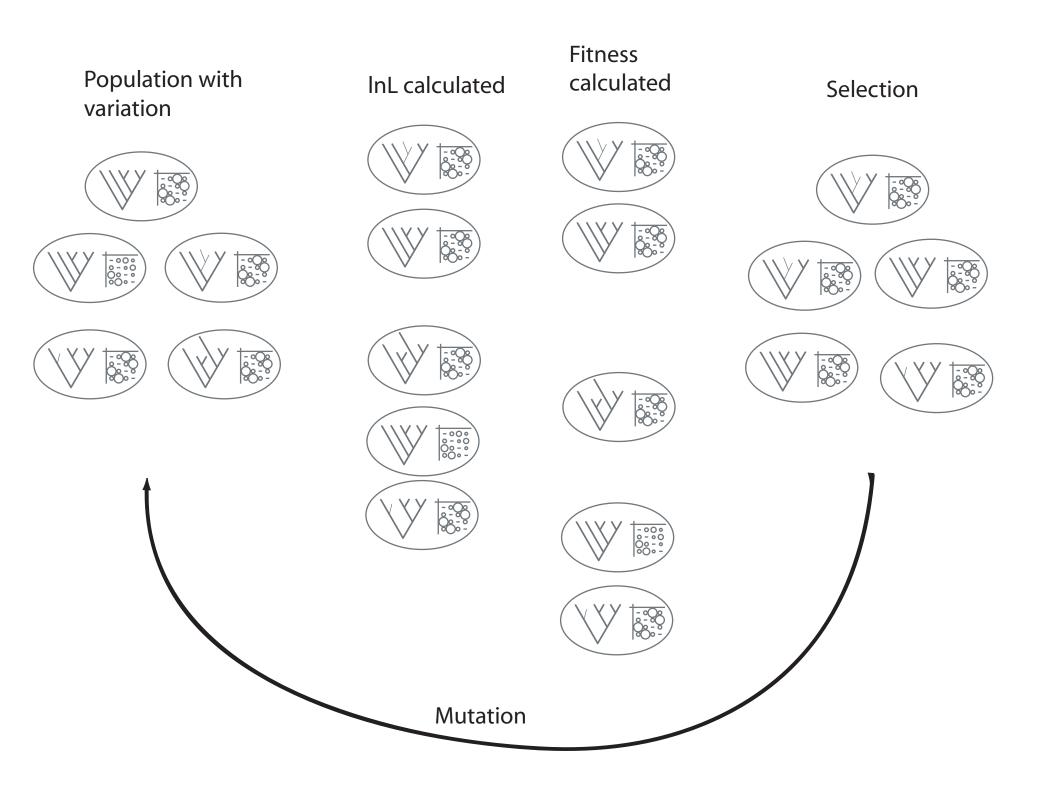






Fitness





Fast tree searching:

- Maximum likelihood RAxML, FastTree. GARLI, phyml, Leaphy
- Distances FastME (balanced minimum evolution); FastTree (profile approximation to balanced minimum evolution); PAUP (other distance-based criteria).
- Parsimony TNT

- 1. The large number of trees make it infeasible to evaluate every tree;
- 2. Intuitive, hill climbing routines often perform well;
- 3. Repeated searching from multiple starting points helps give you a sense of how difficult searching is for your dataset.
- 4. The ease of tree searching is a separate issue from statistical support. Well-supported clades are often easy to find, but we do **not** simply use the repeatability of a trees in independent searches as a measure of support (we'll talk about assessing support tomorrow).

Saitou, N. and Nei, M. (1987). The neighbor-joining method: a new method for reconstructing phylogenetic trees. *Molecular Biology and Evolution*, 4(4):406–425.