Lecture 1.3: Phylogenetic Methods

Popular phylogenetic methods

- 1. Maximum parsimony
- Distance-based methods
- Maximum likelihood
- 4. Bayesian inference

Model-based methods









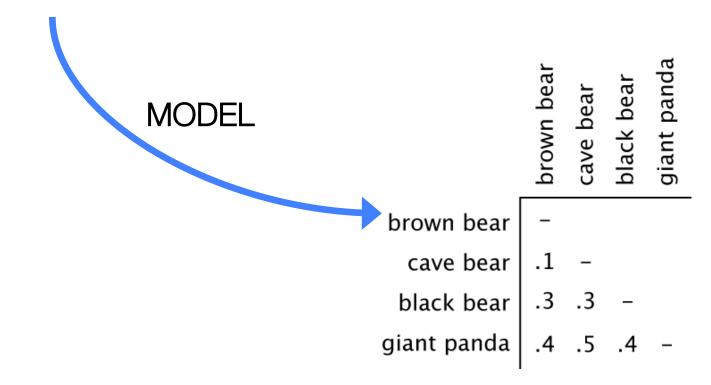




Distance-Based Methods

Distance-based methods

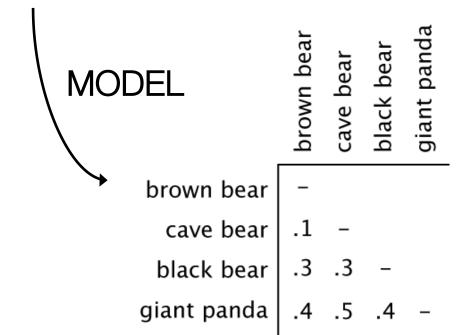
brown bear CGTTAGTACACT cave bear CGATAGTTCACT black bear CGTTAGTTTACC giant panda CATTGGTTTACT

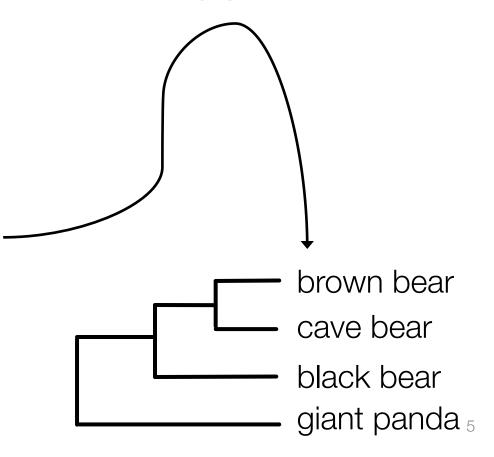


brown bear cave bear black bear

CGTTAGTACACT **CGATAGTTCACT** CGTTAGTTTACC giant panda CATTGGTTTACT

CLUSTERING ALGORITHM



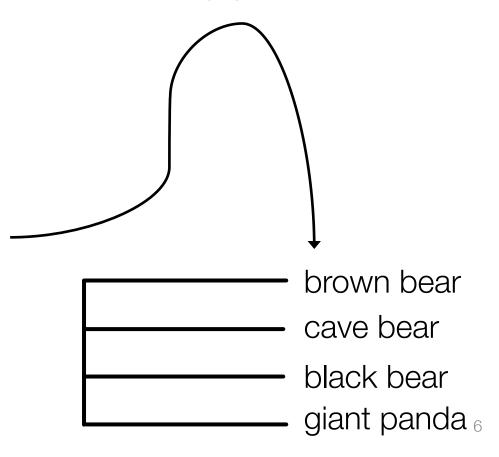


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CLUSTERING ALGORITHM

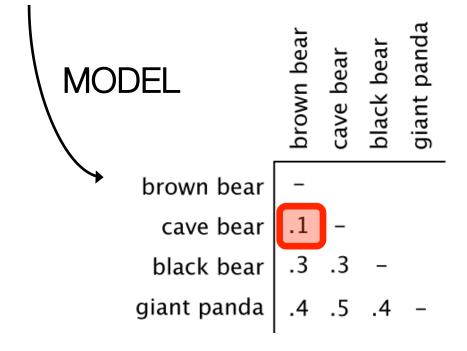
giant panda black bear cave bear MODE brown bear cave bear black bear giant panda

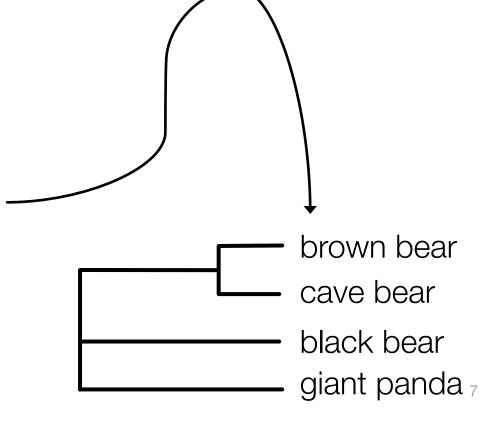


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CLUSTERING ALGORITHM

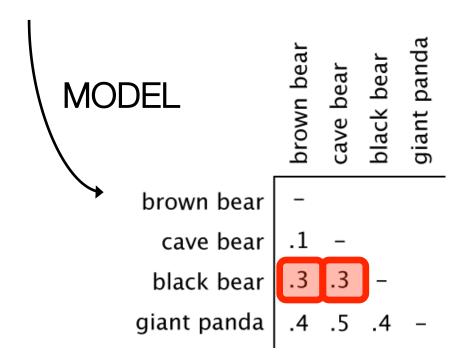


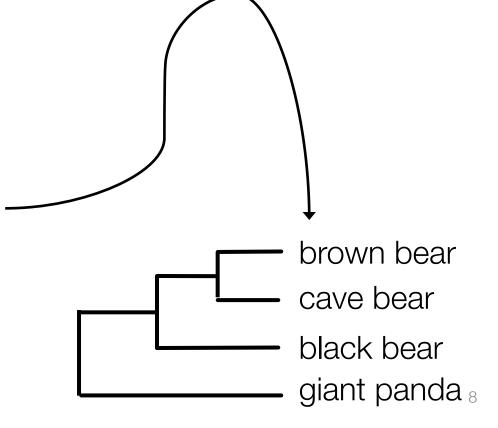


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CLUSTERING ALGORITHM





Distance-based methods

Clustering algorithms

- Unweighted Pair Group Method with Arithmetic Mean (UPGMA)
- Neighbour joining

- Tree searching using optimality criteria
 - Minimum evolution
 - Least-squares inference

Strengths and weaknesses

Strengths

- Very quick method
- Deals with multiple substitutions and long-branch attraction

Weaknesses

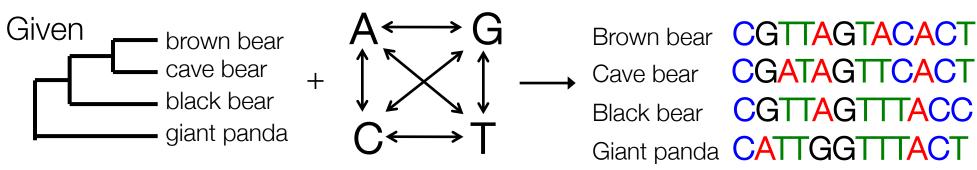
- Does not use all information in alignment
- Loss of information in pairwise comparisons
- Unable to implement sophisticated evolutionary models

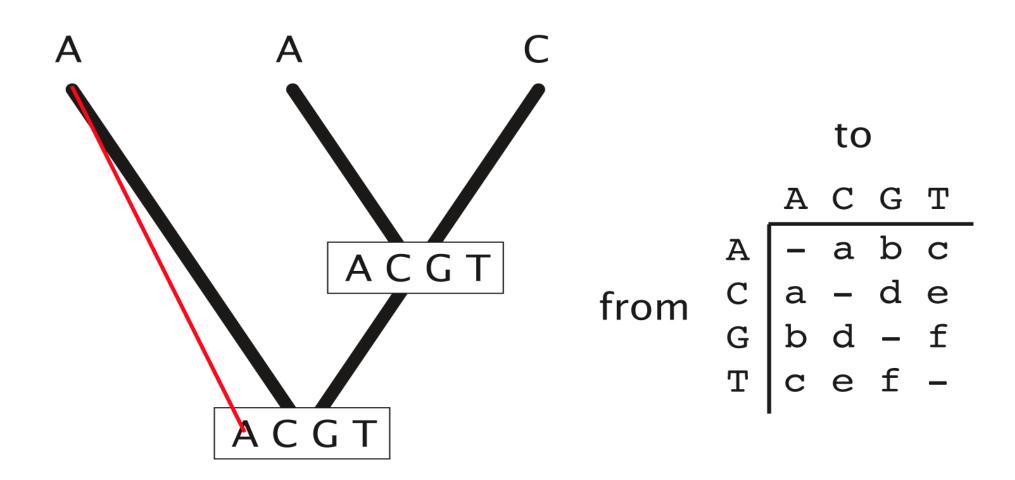
Likelihood of hypothesis H =

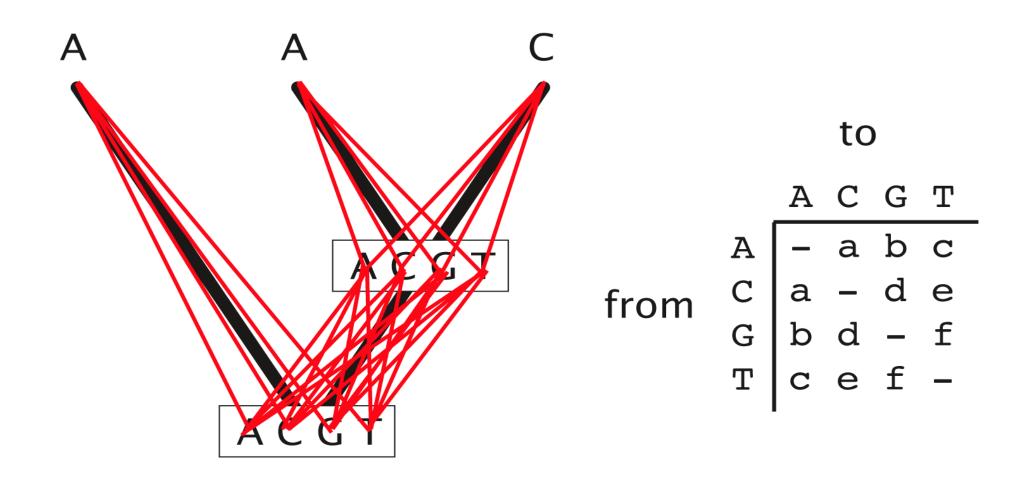


the probability of the data, given the hypothesis

Probability of?







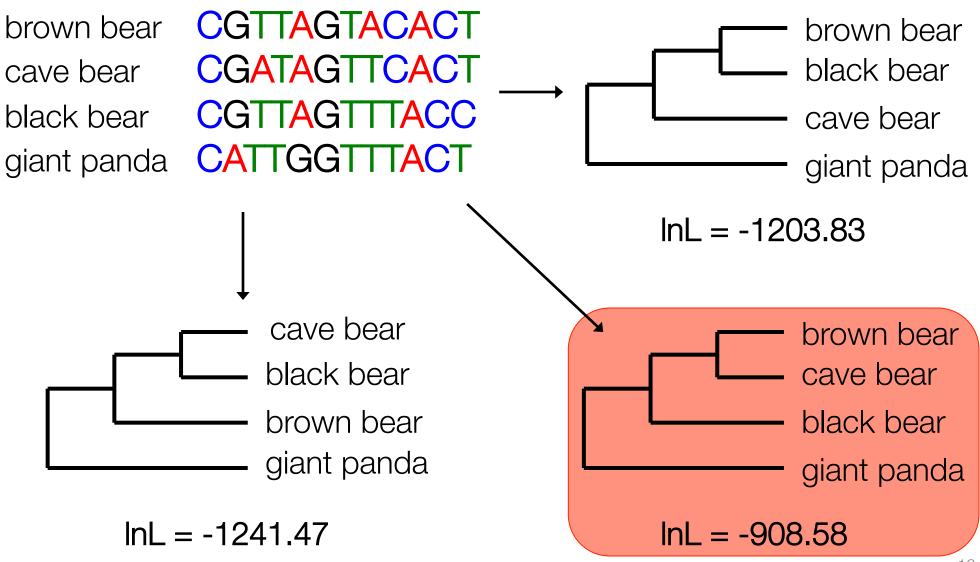
Likelihood = sum of all possible scenarios

Likelihood is multiplied across sites

brown bear cave bear black bear giant panda L₁L₂L₃ ...

CGTTAGTACACT
CGATAGTTCACT
CGTTAGTTTACC
CATTGGTTTACT

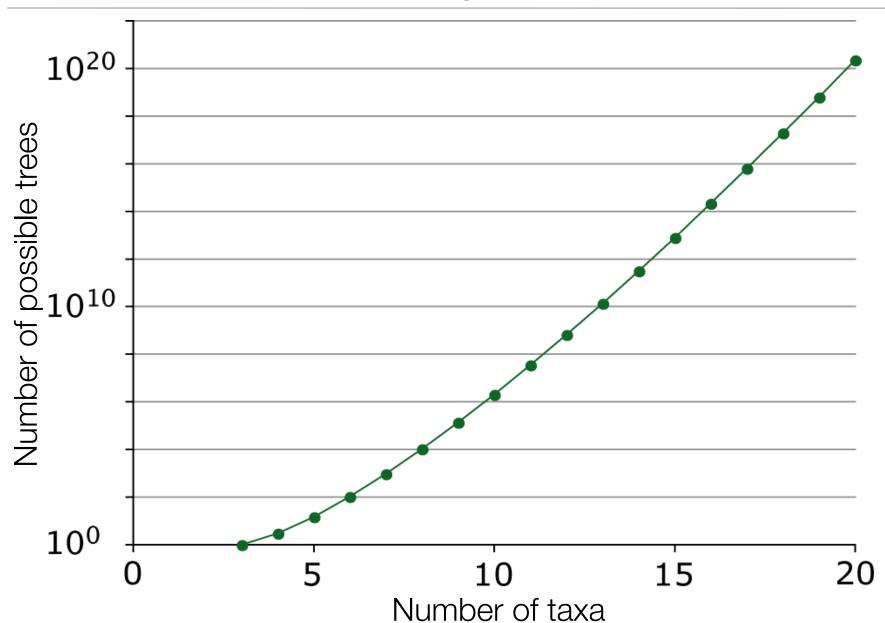
Likelihood values are very small!



Likelihood optimisation

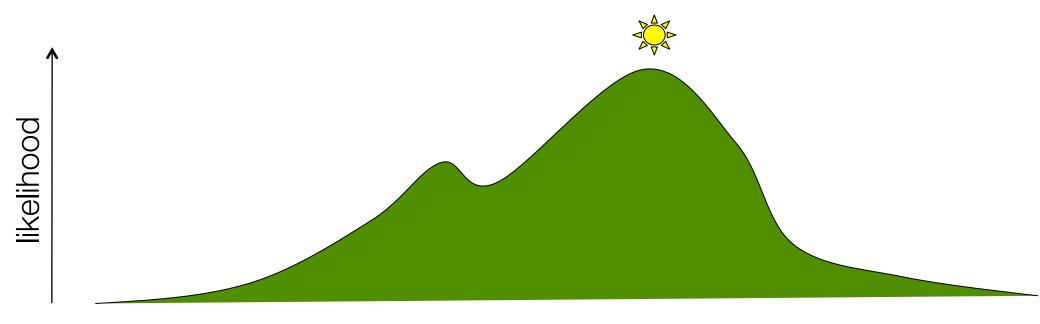
- Search through the space of possible trees and parameter values
- Calculate the likelihood for these
- Find best tree and model parameter values
- Multivariate optimisation

Searching tree space



Heuristic search

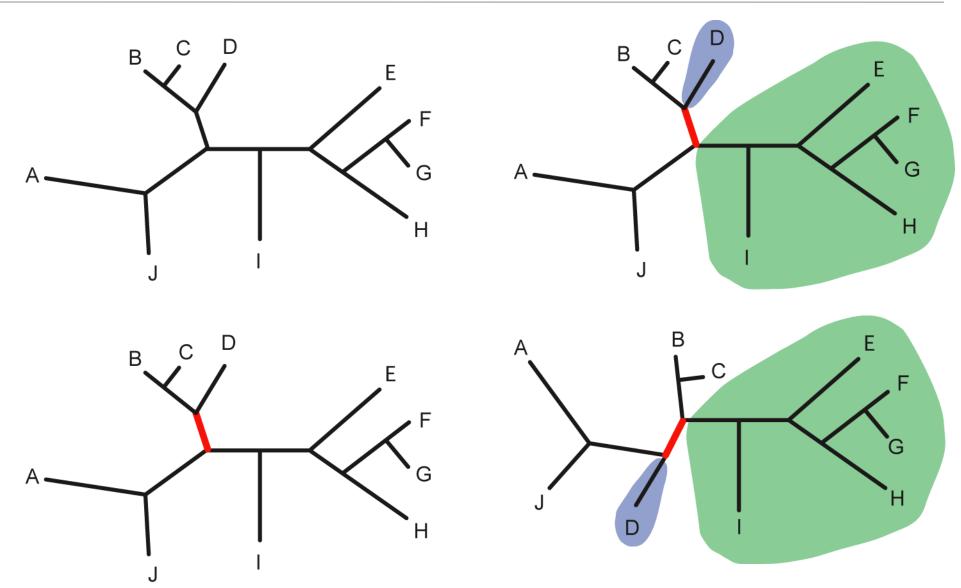
Heuristic search algorithms



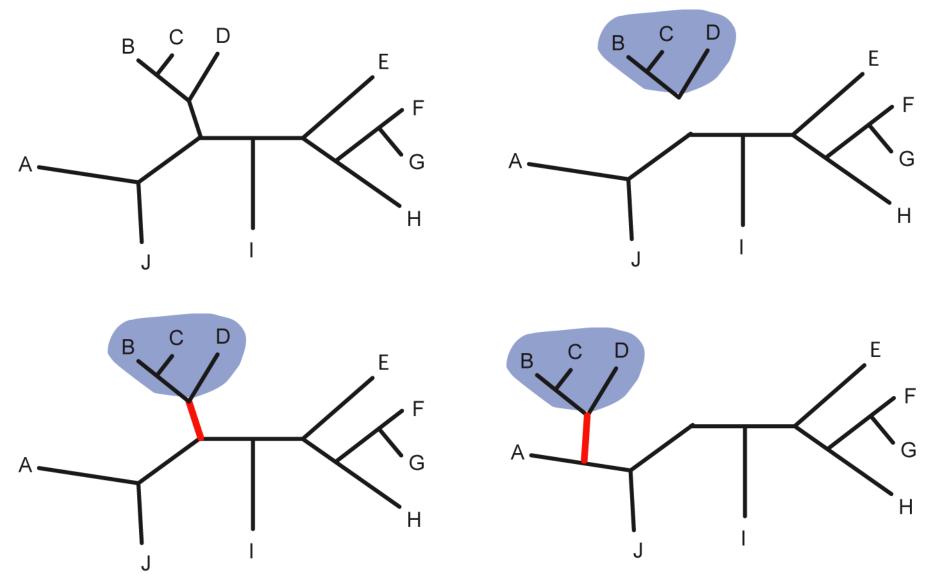
Methods of proposing changes to trees that result in nearby trees:

- Nearest-neighbour interchange (NNI)
- Subtree prune and regraft (SPR)
- Tree bisection and reconnection (TBR)

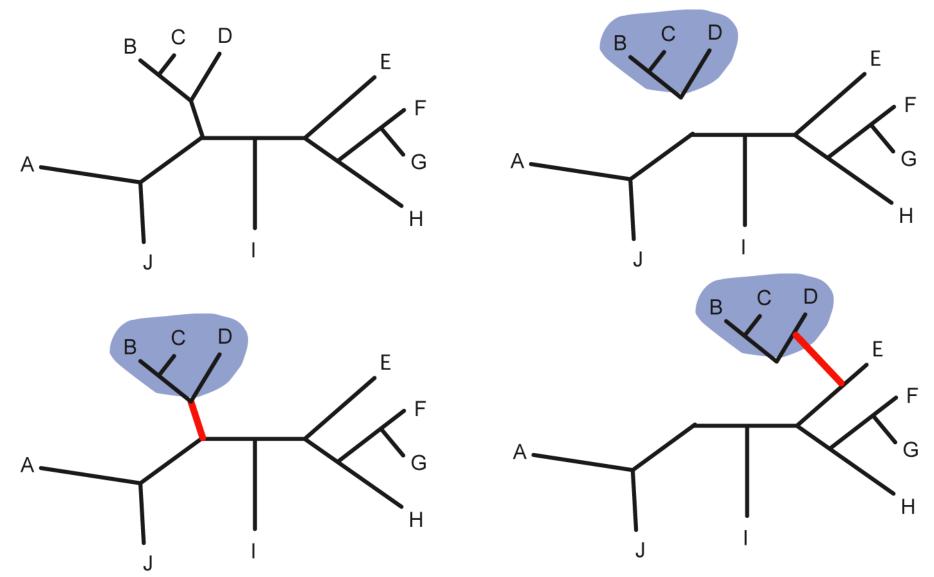
Nearest-neighbour interchange (NNI)



Subtree prune and regraft (SPR)



Tree bisection and reconnection (TBR)

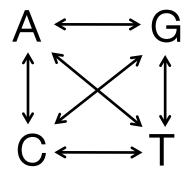


Heuristic search

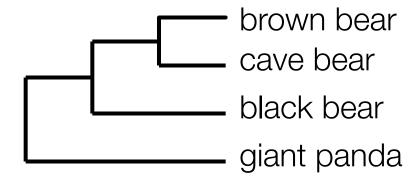
- 1. Pick a starting tree (e.g., NJ or a random tree)
- 2. Use heuristic search to improve model parameters
- 3. Use heuristic search to improve branch lengths
- 4. Use NNI, SPR, and/or TBR to look for a better tree
- 5. Go to step 2

The result

A single set of maximum-likelihood estimates of model parameters



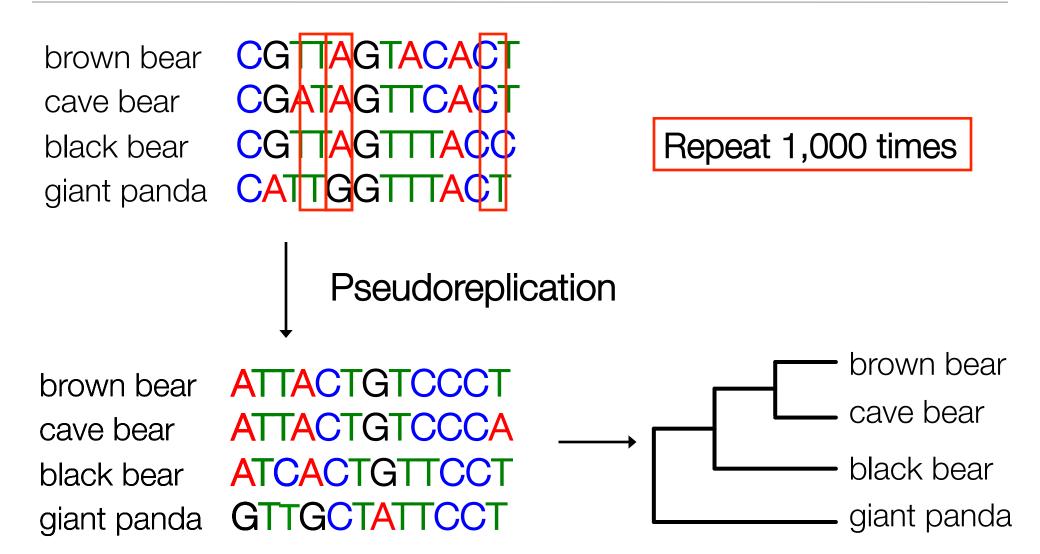
A single maximum-likelihood tree



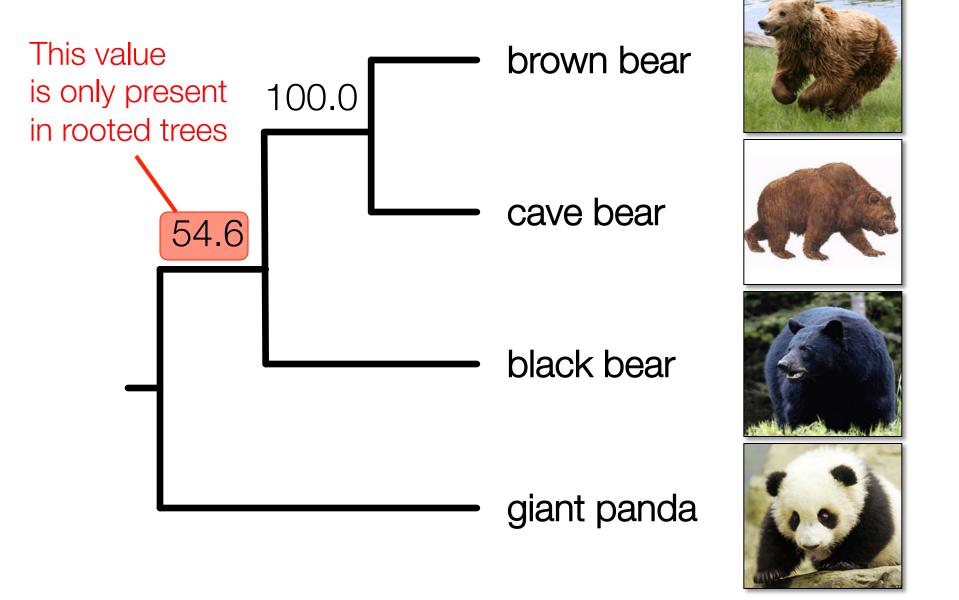
Confidence intervals

- For MLEs of model parameters:
 - Can use the normal approximation (assumes symmetric variance around MLE)
 - 95% confidence interval is: MLE ± (1.96 x stdevMLE)
- We cannot construct a confidence interval for the tree
 - Instead, uncertainty is estimated indirectly using bootstrapping analysis

Bootstrapping



Bootstrapping



Strengths and weaknesses

Strengths

- Rigorous statistical method
- Desirable statistical properties
- Highly robust to violations of assumptions

Weaknesses

- Not feasible to implement very parameter-rich models
- Searching tree-space can be difficult
- Need to rely on heuristic search methods
- Bootstrapping analysis is very slow

Software





PAUP

PhyML



Garli

MEGA



RAxML

Phylogenetic methods

| | Algorithm- based | Optimality criterion | Other |
|--------------------------------|-------------------------------|-----------------------|--------------------|
| No explicit substitution model | | Maximum parsimony | |
| Explicit substitution model | Distance- based methods | Maximum likelihood | Bayesian inference |

Go to Practical 1b: Model selection in MEGA

Go to Practical 1c: Maximum likelihood in PhyML