Lecture 1.4

Phylogenetic Methods

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Popular phylogenetic methods

- 1. Maximum parsimony
- 2. Distance-based methods
- Maximum likelihood

Model-based methods

4. Bayesian inference





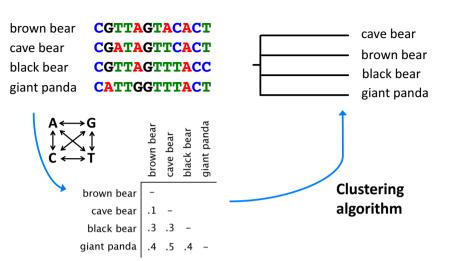




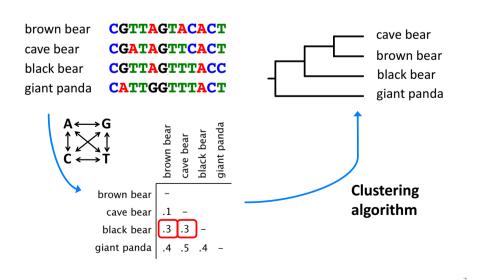
Maximum parsimony brown bear CGTTAGTACACT brown bear **CGATAGTTCACT** cave bear black bear CGTTAGTTTACC cave bear black bear giant panda **CATTGGTTTACT** giant panda 7 steps cave bear brown bear black bear cave bear brown bear black bear giant panda giant panda 7 steps 6 steps

Distance-Based Methods

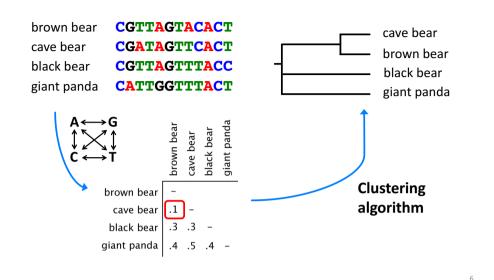
Neighbour joining



Neighbour joining



Neighbour joining



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

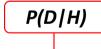
Weaknesses

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

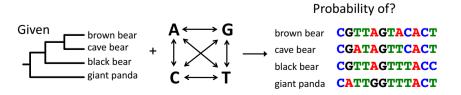
Maximum Likelihood

Maximum likelihood

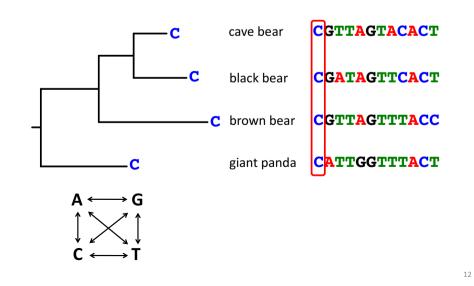
Likelihood of hypothesis *H* =

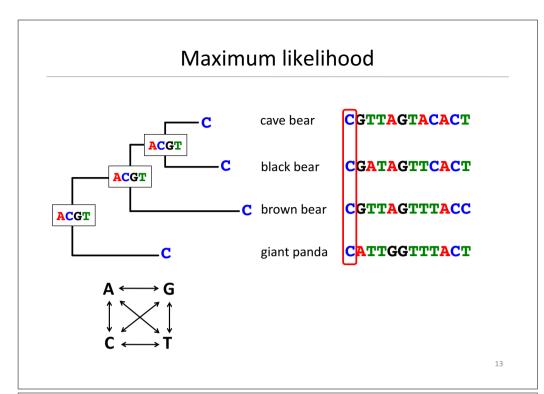


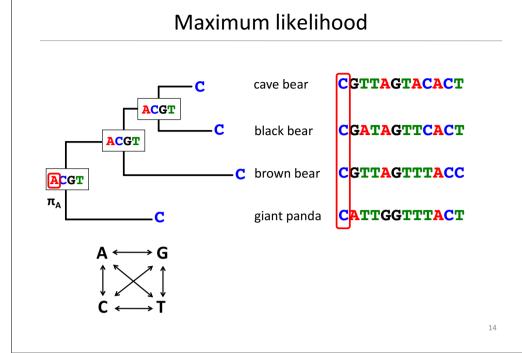
Probability of the data, given the hypothesis

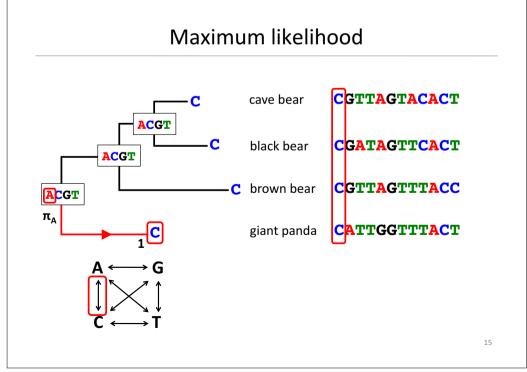


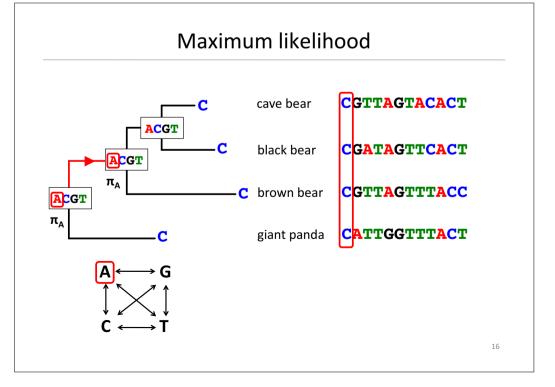
Maximum likelihood

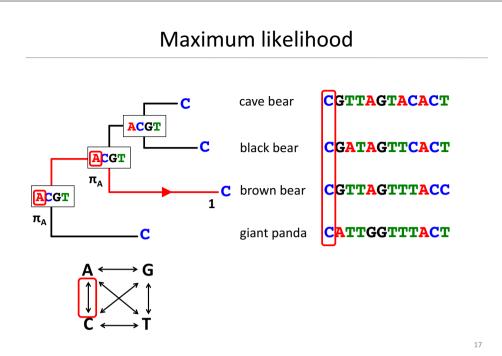


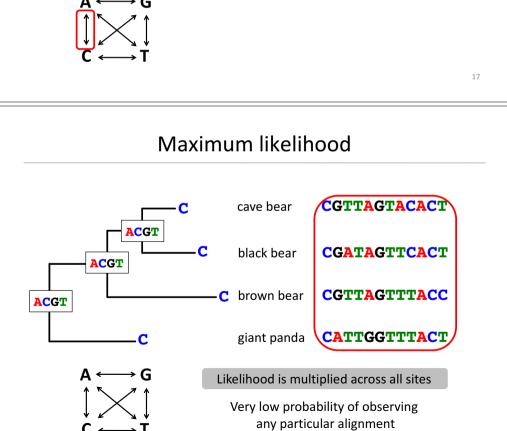


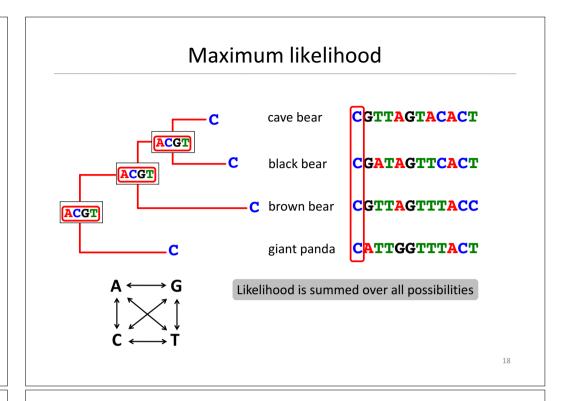


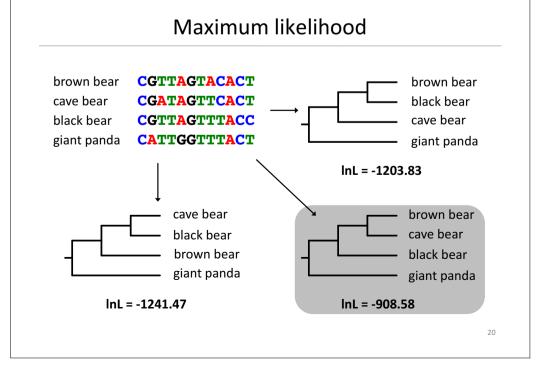






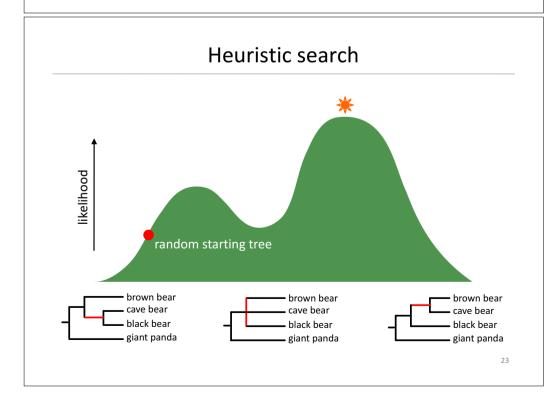


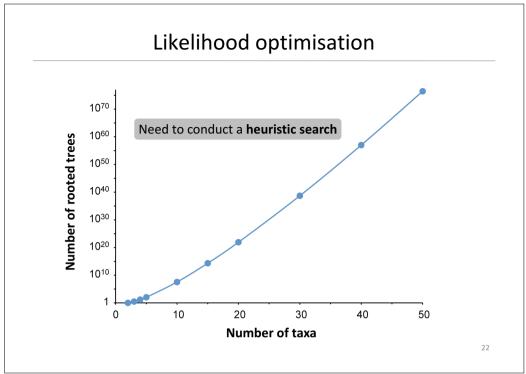


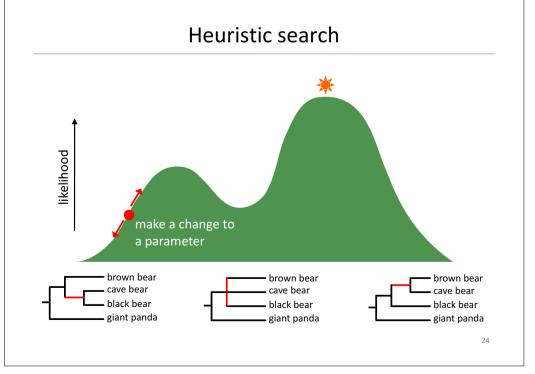


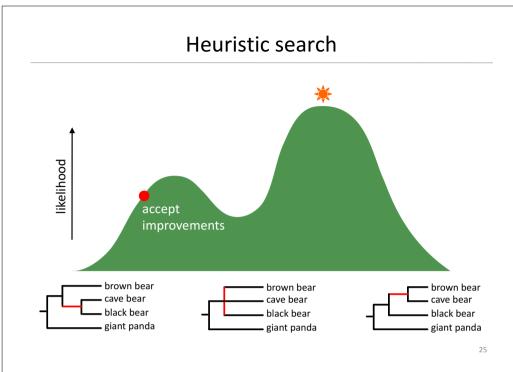
Likelihood optimisation

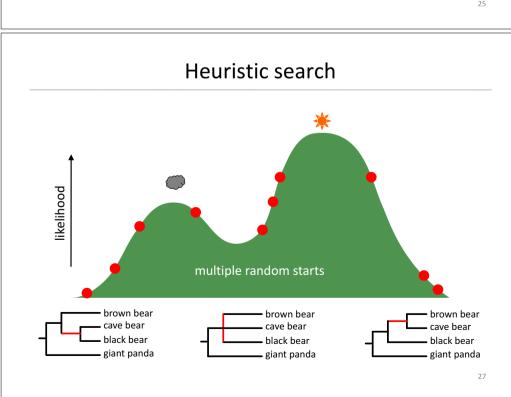
- Search through the space of possible trees and parameter values
- · Calculate the likelihood for these
- Find best tree and model parameter values
- Multivariate optimisation

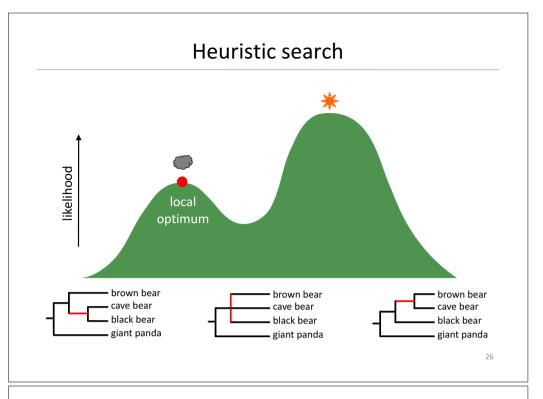


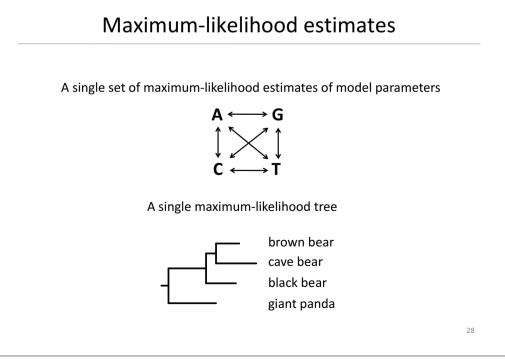












Strengths and weaknesses

Strengths

- · Rigorous statistical method
- Deals with multiple substitutions and long-branch attraction
- · Highly robust to violations of assumptions

Weaknesses

- Not feasible to implement very parameter-rich models
- · Searching tree space can be difficult

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RAxML

- Randomized Axelerated Maximum Likelihood
- · Compile to suit your processor architecture
- Can run sequentially or in parallel
- Rapid bootstrapping (Stamatakis et al. 2008)



Software







PAUI



Garli

MEGA

Mesquite



Jaii

ExaML

- Exascale Maximum Likelihood
- Phylogenetic inference on supercomputers
- New MPI parallelisation approach
- Koslov, Aberer, & Stamatakis (2015) Bioinformatics



Bootstrapping

Nonparametric bootstrap

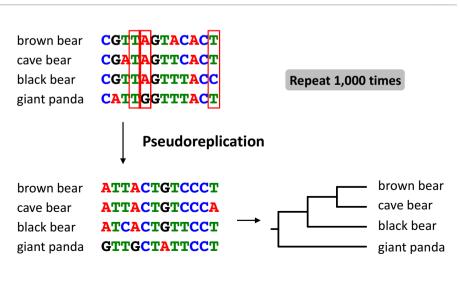
- Uncertainty in the estimate of the tree is inferred indirectly using bootstrapping analysis
- "pull oneself up by one's bootstraps"



- Bootstrapping analysis can be used in various phylogenetic methods:
 - Maximum parsimony
 - Distance-based methods
 - · Maximum likelihood

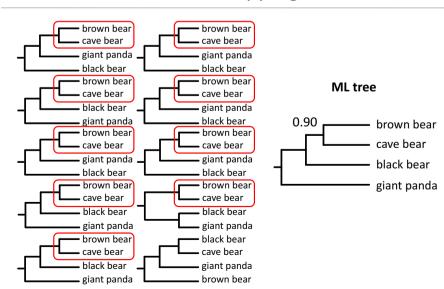
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Bootstrapping

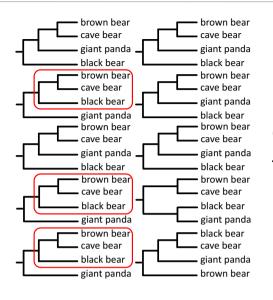


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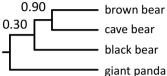
Bootstrapping



Bootstrapping



ML tree

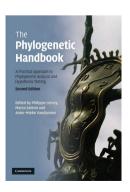


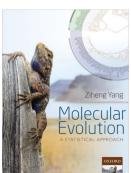
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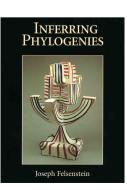
Useful references

Phylogeny estimation and hypothesis testing using maximum likelihood

Huelsenbeck & Crandall (1997) Annu Rev Ecol Syst, 28: 437-466.







Interpreting bootstrap values

• Felsenstein (1985)

bootstrapping provides a confidence interval that contains the phylogeny that would be estimated from repeated sampling of many characters from the underlying set of all characters

- Bootstrap values are measures of repeatability
 - · High when the data set is large
 - · Not meaningful when analysing genome-scale data

Soltis & Soltis (2003) Stat Sci

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Phylogenetic methods

| | Algorithm- based | Optimality criterion | Other |
|---|------------------------|-----------------------|-----------------------|
| No explicit substitution model | Distance-based methods | Maximum parsimony | |
| $ \begin{array}{ccc} A \longleftrightarrow G \\ \uparrow & \downarrow & \uparrow \\ C \longleftrightarrow T \end{array} $ | Distance-based methods | Maximum likelihood | Bayesian inference |

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