Lecture 11 Molecular phylogenetics



Course: Practical Bioinformatics (BIOL 4220)

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Lecture 11 outline

Last time: regex

This time: phylogenetics

- interpreting trees
- tree-thinking
- inferring trees
- inference methods

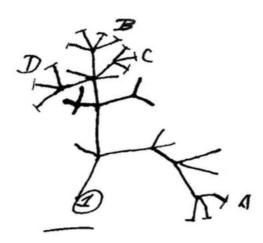
Phylogenetics

Phylogenetics studies the relationships among heritable biological entities (often called **taxa**)

Phylogenies are useful for

- gene annotation
- tracking viral spread
- identifying zoonosis
- reconstructing tumorogenesis
- conservation biology assays
- inferring species relationships

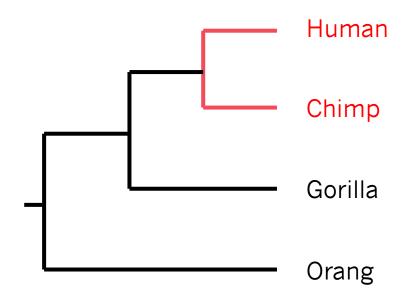
I think



phylogeny sketch by Darwin

Reading a phylogeny

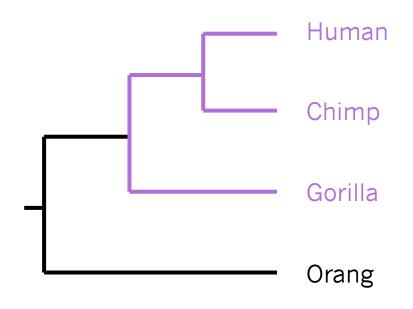
Phylogenetic relationships are hierarchical, and most often represented as bifurcating *trees*



Human and Chimp are more closely related to each other than to Gorilla or Orang

Reading a phylogeny

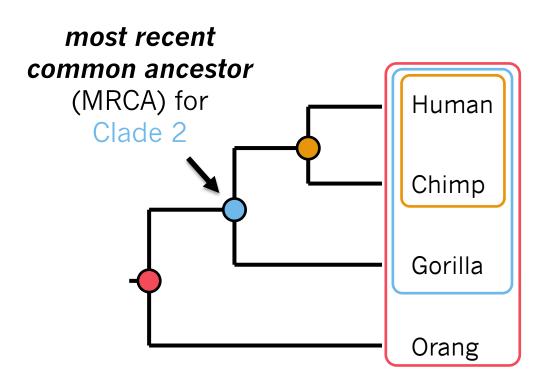
Phylogenetic relationships are hierarchical, and most often represented as bifurcating *trees*



Human, Chimp and Gorilla are more closely related to each other than to Orang

Reading a phylogeny

Taxa that are more closely related to one another, over any other taxa, are called *clades*



Clade 1: H+C

Clade 2: H+C+G

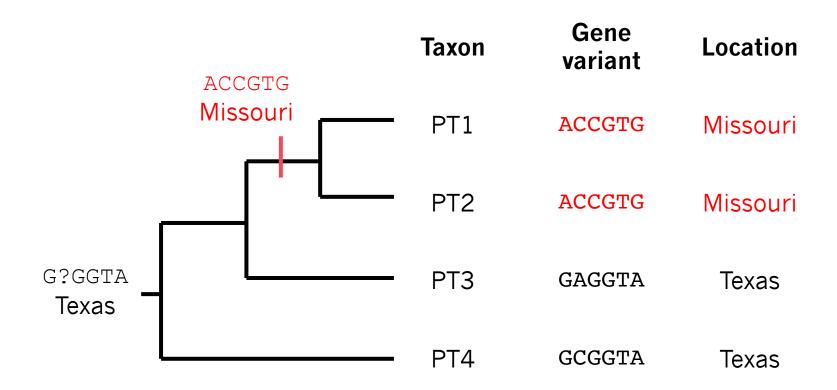
Clade 3: H+C+G+O

"Tree-thinking"

Taxon	Gene variant	Location
PT1	ACCGTG	Missouri
PT2	ACCGTG	Missouri
PT3	GAGGTA	Texas
PT4	GCGGTA	Texas

Four sequences, but no historical context

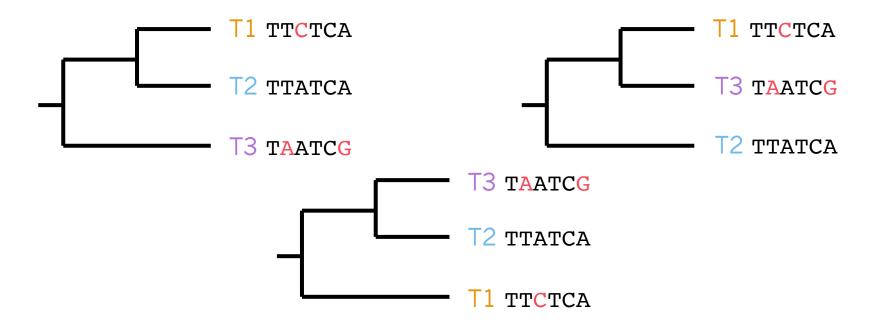
"Tree-thinking"



Phylogeny informs when and where variation arose, which can guide future research

Inferring phylogeny

How are taxa T1, T2, and T3 related?



Which phylogeny generated the observed pattern of molecular variation?

Inferring phylogeny

Phylogenetic inference methods take a matrix of characters (e.g. DNA alignment) as input

Measure how well any possible phylogenetic estimate explains the data matrix pattern by assigning a *cost* to each considered estimate

Methods generally **optimize** the cost to estimate the phylogeny with the lowest cost for the provided data matrix

Tree-space is large

A major challenge: how to efficiently search for trees with optimal scores?

Phylogenetic method types

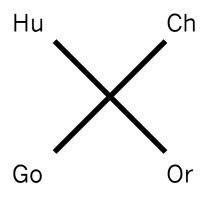
Most methods used to infer phylogenies compute scores based on

- 1. pattern distances (e.g. neighbor joining)
- 2. event counting (*parsimony*)
- 3. event probabilities (*likelihood*)

Method choice often relates to concerns regarding accuracy, speed, scalability, etc.

Neighbor-joining

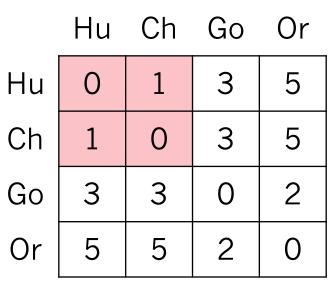
	Hu	Ch	Go	Or
Hu	0	1	3	5
Ch	1	0	3	5
Go	3	3	0	2
Or	5	5	2	0



distance matrix for sequence pairs

Select pairs of taxa with short sequence distances, and join them as neighbors

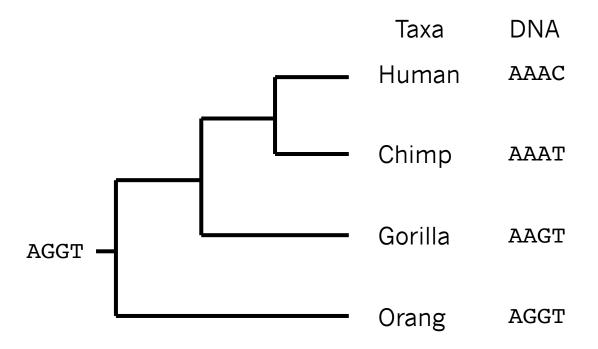
Neighbor-joining

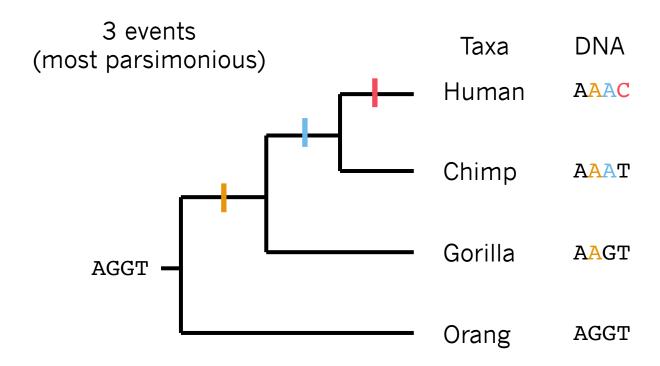


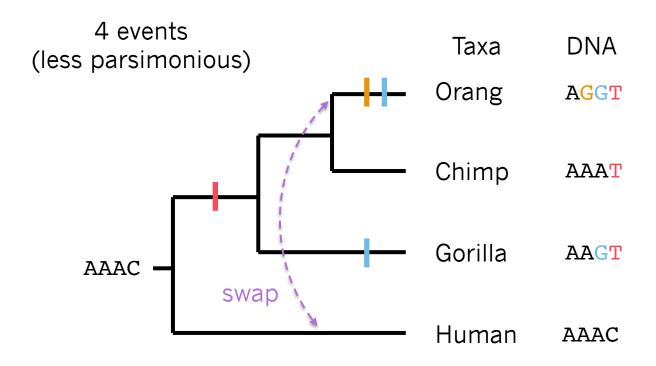
distance matrix for sequence pairs

Hu Ch Hu and Ch form a cluster Or Go Hu Ch Go

Select pairs of taxa with short sequence distances, and join them as neighbors

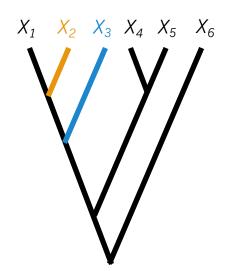


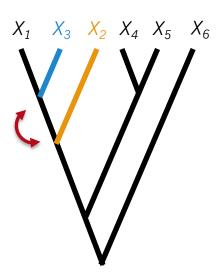




Exploring tree space

Define stochastic "moves" that modify topology, prefer moves that improve tree score

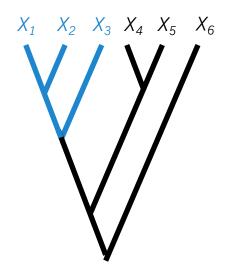


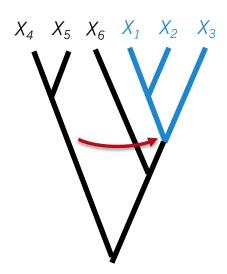


Nearest neighbor interchange (NNI)

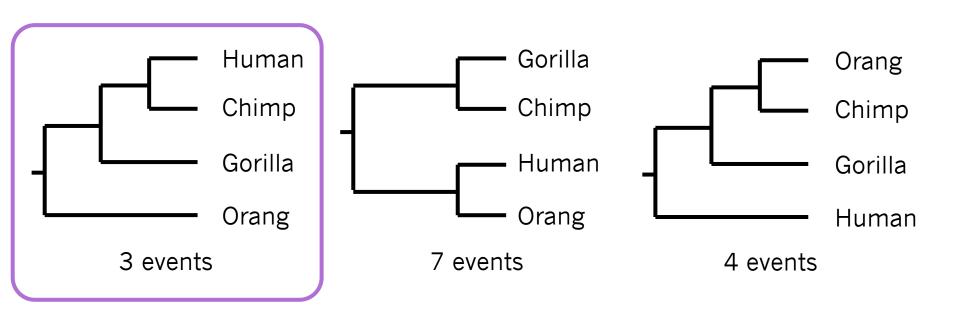
Exploring tree space

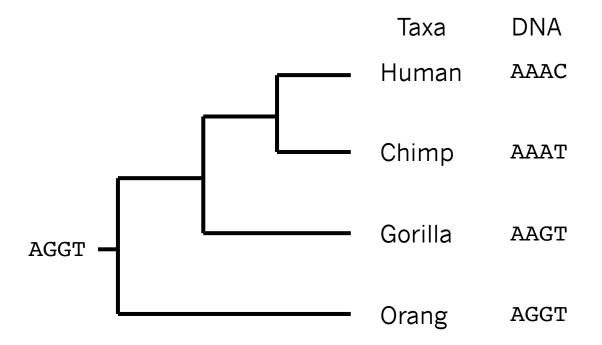
Define stochastic "moves" that modify topology, prefer moves that improve tree score





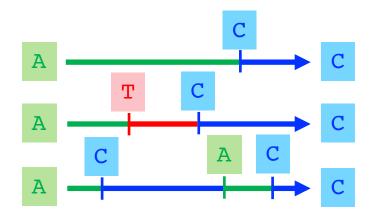
Subtree-prune-regraft (SPR)





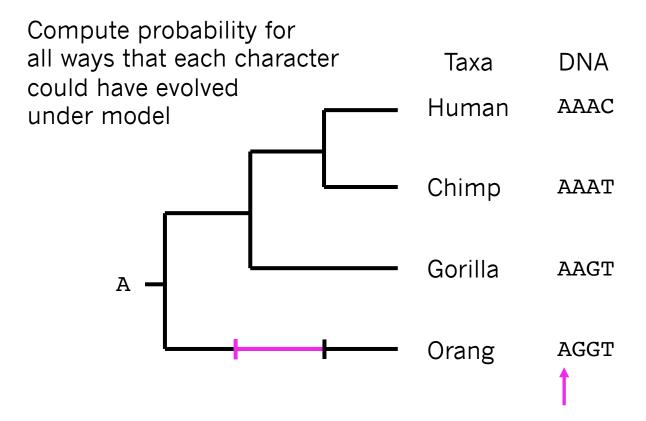
DNA evolution on branch

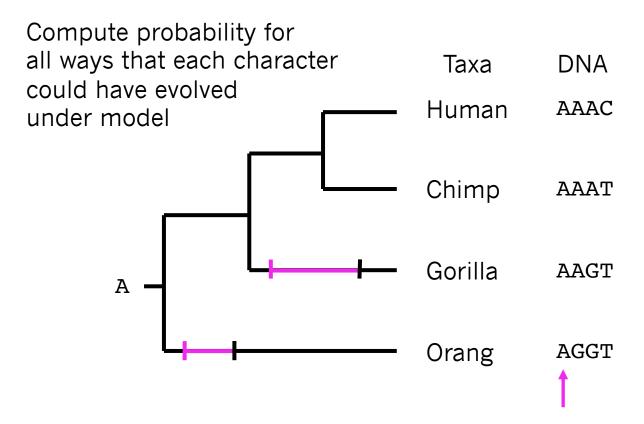
A single DNA site for branch *k* at time *t* can be in one four discrete states: A, C, G, T

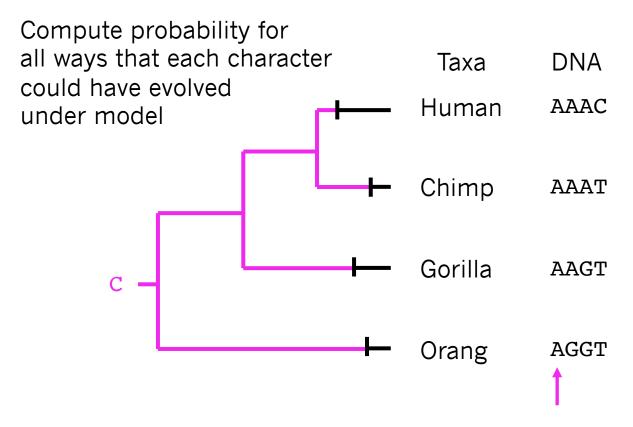


(possible evolutionary histories)

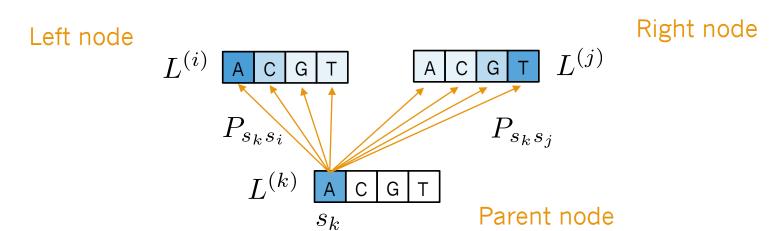
What is the probability that a DNA site in state *i* will end in state *j* after time *t* for branch *k*?







Compute partial likelihood ($L^{(k)}$) for each start state (s_k) against all end states (s_i , s_i)

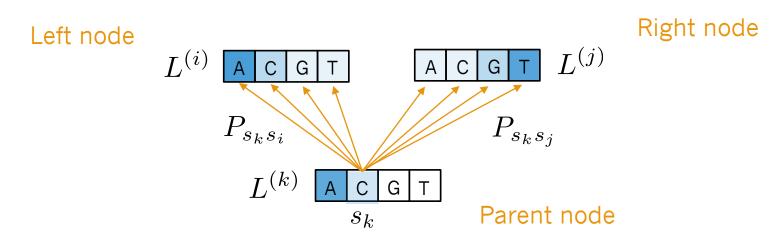


Example for partial likelihood $L^{(k)}$ at node X_k for state $S_k = A$

$$L_A^{(k)} = \left(P_{AA}(t_i) L_A^{(i)} + P_{AC}(t_i) L_C^{(i)} + P_{AG}(t_i) L_G^{(i)} + P_{AT}(t_i) L_T^{(i)} \right)$$

$$\times \left(P_{AA}(t_j) L_A^{(j)} + P_{AC}(t_j) L_C^{(j)} + P_{AG}(t_j) L_G^{(j)} + P_{AT}(t_j) L_T^{(j)} \right)$$

Compute partial likelihood ($L^{(k)}$) for each start state (s_k) against all end states (s_i , s_i)

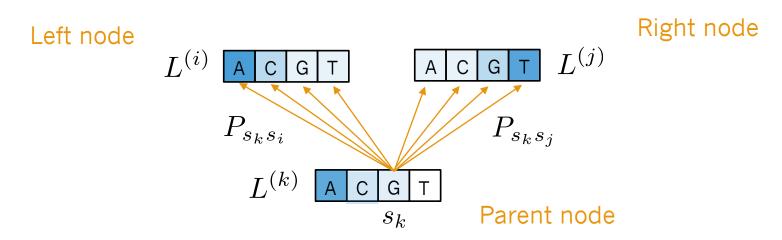


Example for partial likelihood $L^{(k)}$ at node X_k for state $s_k = C$

$$L_C^{(k)} = \left(P_{CA}(t_i) L_A^{(i)} + P_{CC}(t_i) L_C^{(i)} + P_{CG}(t_i) L_G^{(i)} + P_{CT}(t_i) L_T^{(i)} \right)$$

$$\times \left(P_{CA}(t_j) L_A^{(j)} + P_{CC}(t_j) L_C^{(j)} + P_{CG}(t_j) L_G^{(j)} + P_{CT}(t_j) L_T^{(j)} \right)$$

Compute partial likelihood ($L^{(k)}$) for each start state (s_k) against all end states (s_i , s_i)

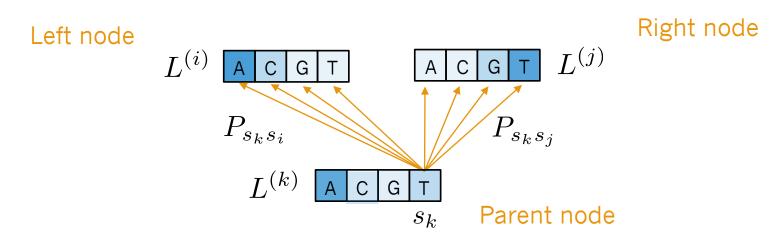


Example for partial likelihood $L^{(k)}$ at node X_k for state $s_k = G$

$$L_G^{(k)} = \left(P_{GA}(t_i) L_A^{(i)} + P_{GC}(t_i) L_C^{(i)} + P_{GG}(t_i) L_G^{(i)} + P_{GT}(t_i) L_T^{(i)} \right)$$

$$\times \left(P_{GA}(t_j) L_A^{(j)} + P_{GC}(t_j) L_C^{(j)} + P_{GG}(t_j) L_G^{(j)} + P_{GT}(t_j) L_T^{(j)} \right)$$

Compute partial likelihood ($L^{(k)}$) for each start state (s_k) against all end states (s_i , s_i)

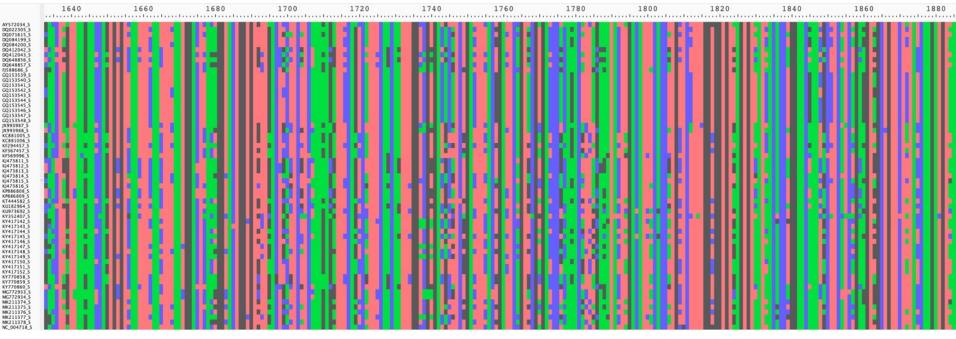


Example for partial likelihood $L^{(k)}$ at node X_k for state $s_k = \mathbf{T}$

$$L_T^{(k)} = \left(P_{TA}(t_i) L_A^{(i)} + P_{TC}(t_i) L_C^{(i)} + P_{TG}(t_i) L_G^{(i)} + P_{TT}(t_i) L_T^{(i)} \right)$$

$$\times \left(P_{TA}(t_j) L_A^{(j)} + P_{TC}(t_j) L_C^{(j)} + P_{TG}(t_j) L_G^{(j)} + P_{TT}(t_j) L_T^{(j)} \right)$$

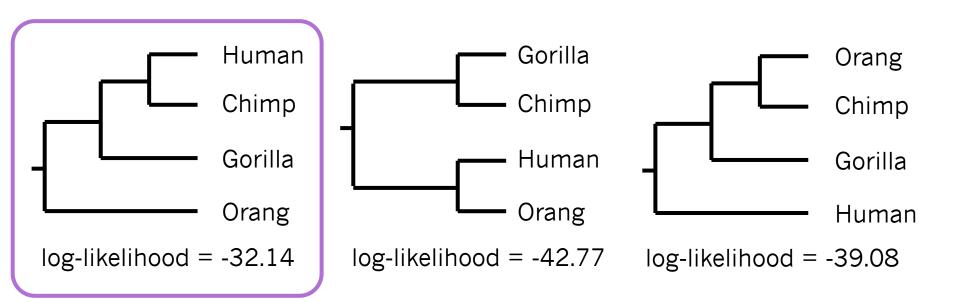
Phylogenetic likelihood





What is the probability of observing a set of genetic variants for a given underlying tree?

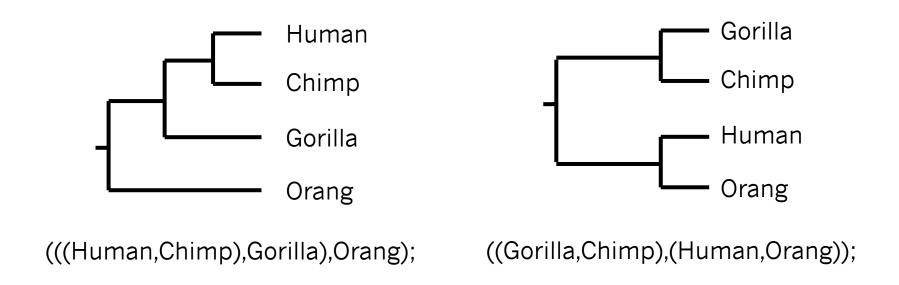
Prob(all sites | tree) = Prob(site 1 | tree) x ... x Prob(site N | tree)



Method comparison

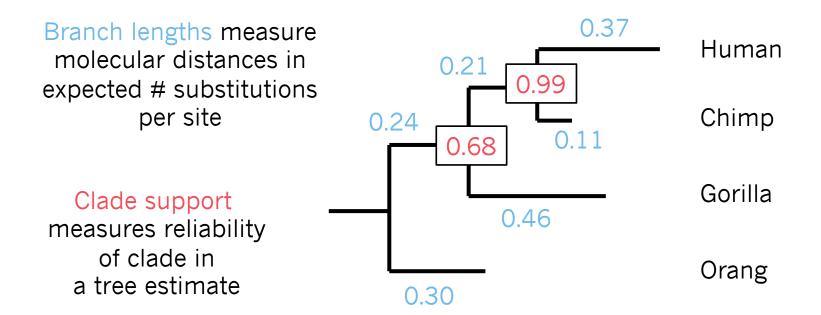
Method	Pros	Cons
Neighbor-joining	Extremely fast Scalable	Does not use evolutionary events to infer tree
Parsimony	Intuitive Fairly fast	Assumes change is rare; Unclear how to choose "costs" for events
Likelihood	Most accurate Most realistic Can simulate data	Slower More complex theory + algorithms

Newick strings



Taxa in parentheses define clades; commas define divergence events

Newick strings



(((Human: 0.37, Chimp: 0.11) 0.99: 0.21, Gorilla: 0.46) 0.68: 0.24, Orang: 0.30);

Overview for Lab 11