

## PARSIMONY

## Parsimony

Consider the sequences

$S_1$ : A A T G

L

$S_2$ : A G T G

S

$S_3$ : A G T C

S

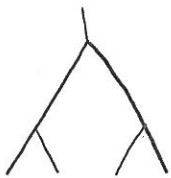
$S_4$ : A A T C

L

DNA

morphological

Which of the 4 (out of  $(2n-3)!! = 5!! = 15$ ) rooted trees, do we prefer?  
and why?



$S_1$   $S_2$   $S_3$   $S_4$

PS = 5



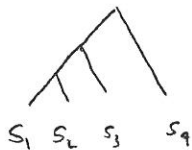
$S_1$   $S_4$   $S_2$   $S_3$

PS = 5



$S_1$   $S_3$   $S_4$   $S_2$

PS = 6



$S_1$   $S_2$   $S_3$   $S_4$

These numbers are called the **PARSIMONY SCORES** of  $T$ ,  $PS(T)$

Subtle point: Parsimony scores are really counts for unrooted trees.

For example, trees 1 and 4 are rooted versions of  $S_1$   $S_2$   $S_3$   $S_4$  and  $PS(T_1) = PS(T_4)$

The three other rooted versions of  $S_1$   $S_2$   $S_3$   $S_4$  would also have the same PS.

Rooting just provides a way (= algorithm) for computing PS.

These numbers are known as the **PARSIMONY SCORES** of  $T$

Theorem:  $ps(T^P) = ps(T)$

i.e. the parsimony score of a rooted tree

= " " " " its unrooted version

$\Rightarrow$  **PARSIMONY SCORES** compare unrooted trees only

The underlying principle: The best tree (or trees in the case of ties) is the one requiring the minimal amount of changes: (substitutions/<sup>state</sup> changes)

Def: A column in aligned sequences is called a **CHARACTER** (usually denoted by  $X$ ) and a character can take on any of  $S$  **STATES**

Eg: DNA character  $S=4$       morphological char  $S=2$

Data:

$X_1 \ X_2 \ X_3 \ \dots$

[ ] [ ] [ ]

How should we compute parsimony scores?  
 $\downarrow$   
 (or have a computer)

Do alg development + Eg Simultaneously

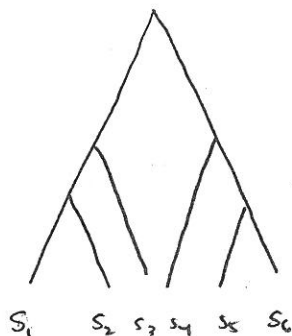
3.

Fix a <sup>binary</sup> tree  $T$

Alg.

Sum over all characters  $X_i, i=1, \dots, m$

find  $PS_{X_i}(T)$  by



A A  
A A  
T A  
T G  
G G  
C C

0. If unrooted, arbitrarily root tree
1. place  $X_i$  at leaves of tree

2. starting at the leaves,

label each ancestral node by

- the union of the two sets below if they are disjoint  
in which case the parsimony count is augmented by 1
- the intersection of the two state sets below if they share etc  
(and do not augment the parsimony count)

3. continue up tree until reach root      Return  $PS_{X_i}(T)$

Comments:

1. This is called the FITCH-HARTIGAN ALGORITHM

and it does compute the  $ps(T)$

Thm: FH count on  $T = ps(T)$

2. computer scientists would call this the SMALL-PARSIMONY PROBLEM

(fix one single tree and compute its  $ps$ )

How complex is this? If  $T$  is an  $X$ -tree with  $n = |X|$  leaves

Roughly,  $n$  internal nodes to visit  
each node might have any of  $s$  states and  
must loop over  $m$  characters

$$\Rightarrow \approx nsm \equiv \Theta(nsm) \quad \text{"lg-}\Theta\text{"}$$

$$\Rightarrow \text{time is } \leq (\text{constant}) \times (nsm)$$

3. The LARGE-PARSIMONY PROBLEM is the one of interest

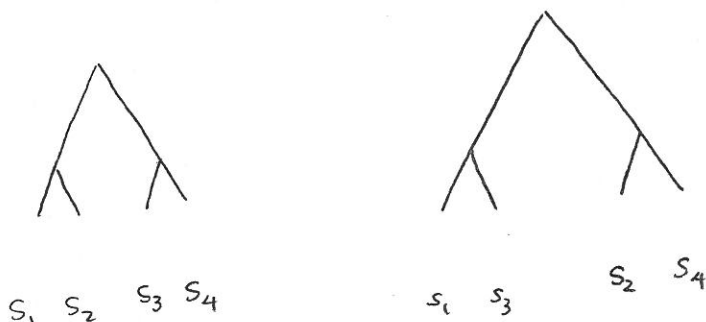
Search over all  $(2n-5)!!$  unrooted trees

compute the parsimony score for each

return those with best score

UAF grad Ron Graham + Foulds proved this is NP-hard

4. Ways to speed this up



$S_1$ :	A	A	A	A	A	A
$S_2$ :	A	G	C	C	A	A
$S_3$ :	A	A	T	C	C	T
$S_4$ :	A	A	G	T	C	T

Remove the constant sites or,  
more generally, the non-informative  
sites

Defn: If  $X$  is a character, with  $S$  states, then a PARSIMONY-INFORMATIVE CHARACTER in which 2 states appear at least twice

A	A	A	S	S
A	A	C	L	S
T	A	G	L	L
T	T	T	L	L

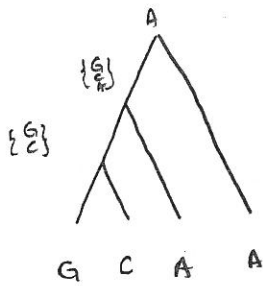
To speed up parsimony, preprocess to remove non-informative sites

recognize that all <sup>informative</sup> patterns,  $x$  contribute equally to  
for instance,  $x$  parsimony count  
 $y$   
 $y$

i.e. particular states don't matter, only pattern of states.

## 5. Reconstructing ancestral states

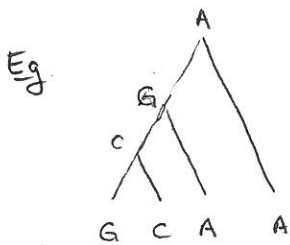
(Beware FH.)



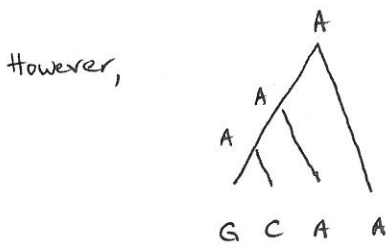
②

A EXTENSION  $\tilde{X}$  of  $X$  is a assignment of states to the internal nodes of  $T$  consistent with  $X$  at the leaves.

$\tilde{X}$  is minimal, if it computes  $pr_X(T)$



$\tilde{X}$  is minimal ext. of  $X$  as is



This extension also gives rise to  $pr = 2$ , but does not arise from FH

extensions  
Bad News: Not all labellings that give rise to  $pr$  for  $X_{onT}$  arise from FH algorithm

Good news: ② You can get some by a second pass down the tree.

Modify FH to get all ...

or use Sankoff algorithm

- ① The state set assigned to  $p$  by FH is exactly the state-set for any minimal ext  $\tilde{X}$  of character  $X$

## WEIGHTED PARSIMONY or the Sankoff algorithm

Underlying idea:

not all characters are equal

→ Character weighting

not all state changes are equal

→ Sankoff alg.

Character-weighting:

CODONS:

2 3 1 2 3 1

→  $x_1 x_2 x_3$ 

$$PS(T) = 2 PS_{x_1}(T) + PS_{x_3}(T) + PS_{x_4}(T)$$

twice as important

Any scheme for character-weighting is legitimate  
if coefficients = weights  $> 0$

In practice, software usually wants integer  
weights

Weighting state changes.

A transition is a state change

 $A \leftrightarrow G, C \leftrightarrow T$ 

transversion

between purines and pyrimidines

 $\{A, G\} \leftrightarrow \{C, T\}$ 

Empirically, transitions are observed much more frequently than transversions

FH all state changes are weighted equally

Perhaps better: introduce a WEIGHT MATRIX, a COST MATRIX, a STEPMATRIX

Eg.  $W =$ 

from

	to			
	A	G	C	T
A	0	1	2	2
G	1	0	2	2
C	2	2	0	1
T	2	2	1	0

Comments: this weight  
matrix is symmetric, but  
this is not necessary

We will use  $w_{ij}$  = cost  
of changing from  
state  $i$  to state  $j$ .

user-defined

Indeed, the weight matrix is <sup>1</sup> judgment on the cost of changes

$W = \text{transversions cost } 2 \times \text{transitions}$

Alternative:  $\chi$  3-state character

	$s_1$	$s_2$	$s_3$
$s_1$	0	1	2
$s_2$	$\infty$	0	1
$s_3$	$\infty$	$\infty$	0

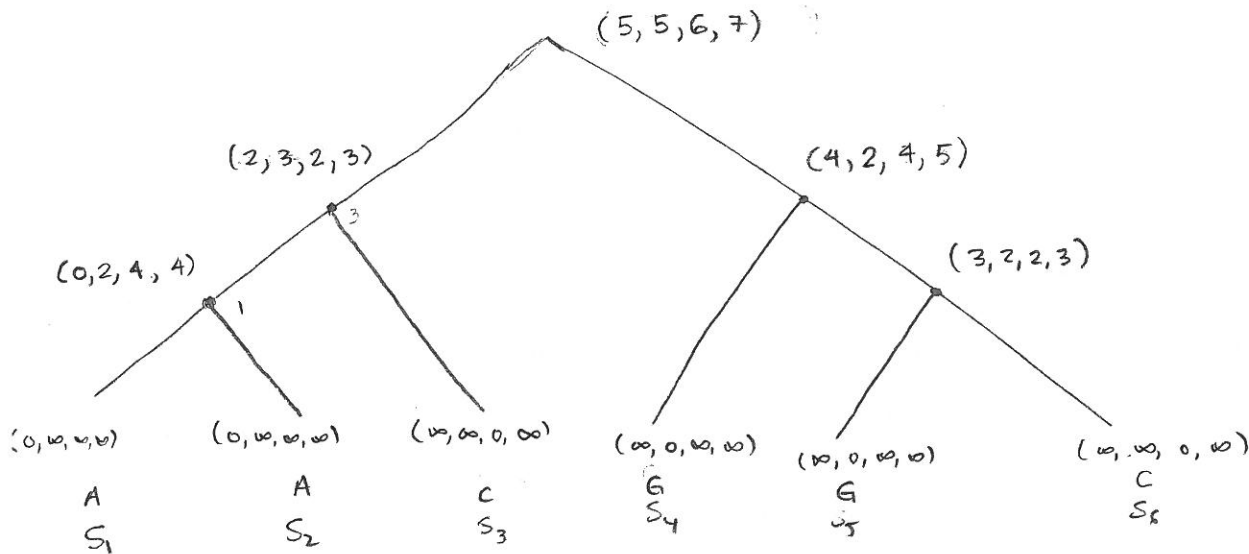


See for eg  
PAUP\* manual on web  
pp 16-20

The Sankoff algorithm for weighted-parsimony by example:  
with transversions twice the cost of transition.

$W =$  to:

from



Main idea: starting at the leaves, and moving up toward root  $\rho$

Place a  $s = 4$  element vector with each entry containing the minimal cost of labelling that node with state  $s_i$

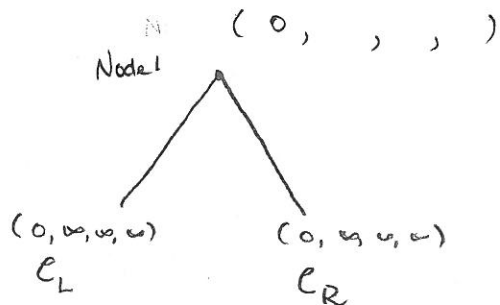
For this particular example, we order the vectors



purines first, pyrimidines 2<sup>nd</sup>

At each tip, form its count vector by placing a 0 in the  $i$ th slot if the tip is in state  $i$ ; or  $\infty$  elsewhere.

Move "up" through the tree focusing on an internal node and its immediate children  $e_L + e_R$



Use the two children and weights  $w_{ij}$  to compute the minimal count vector.

Eg:

Node 1:

A

Cost of changes

min cost at  $e_L$

→

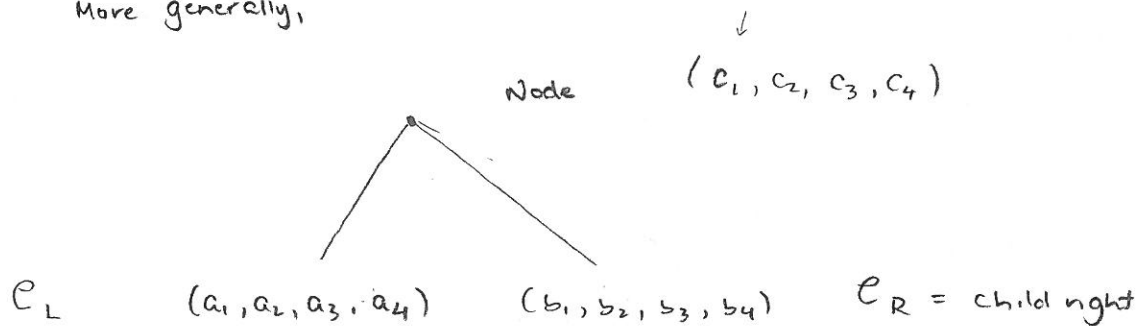
0 1 2 2  
(0, ∞, ∞, ∞)

(0, ∞, ∞, ∞)

0 + 0

minimize { (cost of change  $a \rightarrow j$ ) + (cost of  $j$  at  $e_L$ ) } +

More generally,



$$\min \left\{ \begin{array}{l} \text{Cost of } A \rightarrow j \\ \text{Cost of } j \\ \text{at } e_L \end{array} \right\} + \min \left\{ \begin{array}{l} \text{Cost of } A \rightarrow k \\ \text{Cost of } e_R(k) \end{array} \right\}$$

$$= \min \{ w_{1j} + a_j \} + \min \{ w_{1k} + b_k \}$$



Sankoff algorithm to compute the parsimony score for an s-state character

$X$  on  $T$  with weight matrix  $W$

i.e. compute  $PS_X(T)$  using weights  $W = (w_{ij})$

0. Arbitrarily root  $T$  to get  $T^P$

1. At each tip of the tree, place an s-element vector

If tip is in state  $i$ , place a 0 in the  $i$ -th position

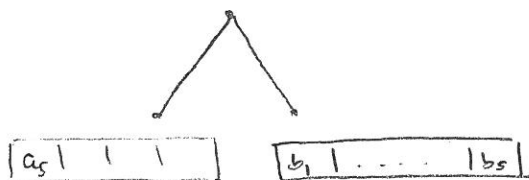
Place an  $\infty$  elsewhere.

2. Move up the tree through all internal nodes.

At each internal node, place an s-element vector with entries

$a_1 | a_2 | \dots | a_s$

computed as follows:



$$\text{Let } a_i = \min_{\substack{\text{states} \\ 1, 2, \dots, s}} \{ w_{ij} + a_j \} + \min_{j \in S} \{ w_{ij} + b_j \}$$

states  
1, 2, ..., s

↑  
cost of state  
change  $i \rightarrow j$

+ cost node in state  $j$   
left child

3. When you reach the root  $P$ ,

Consider its vector  
↓  
Count

$a_1 | a_2 | \dots | a_s$

Then  $PS_X(T) = \text{minimum entry.}$

Theorem: The Sankoff algorithm does compute  $PS_X(T)$  with weights  $W$

Comments: This is an example of a "dynamic programming algorithm"

One advantage: all minimal exts are computable by examining the partial count vectors.

• If the weight matrix  $W$  is symmetric, then ps evaluates unrooted trees  
 is not symmetric, " " is a measure on rooted trees

Reflections:

1. There are many other types of parsimony possible. See Chap 7 Felsenstein  
 (Good project possibly?)

2. Strengths

vs.

Weaknesses

heuristics are needed

Since we can't explore  
 tree space  $(2n-5)!!$

Change is rare is  
 reasonable.

What if taxa have evolved  
 over a "long" period of  
 time? Is it reasonable  
 to avoid consideration of  
 multiple changes?

$L \rightarrow M \rightarrow S$   
 parallel evolution?

What is the meaning of  
 the parsimony scores?

For morphological data,  
 this seems a good choice.

What is the meaning of  
 $w_{ij}$ ?