

# Identifiability of Models from Parsimony-Informative Pattern Frequencies



John A. Rhodes  
University of Alaska  
Fairbanks 

---

June 10, 2008  
MIEP

Joint work with

Elizabeth Allman (UAF)

Mark Holder (U Kansas)

Thanks to the Isaac Newton Institute

## I: Parsimony-informative models:

- Variants of standard Markov substitution models on trees where *only* parsimony-informative patterns are observed
- Useful for phenotypic datasets — acquisition bias prevents appropriate sampling of non-informative character patterns (e.g., all equal, all different)

- Despite shortcomings of simple models for phenotypic datasets, statistical approaches such as ML, Bayesian inference might still be preferable to parsimony
- Model proposed by P. Lewis (2001) omits constant patterns; model of Ronquist–Hulsenbeck (2004?) omits parsimony-noninformative patterns; used for combined analysis of sequence and morphological data by Nylander–Ronquist–Hulsenbeck–Nieves-Aldrey (2004)

For this talk focus on

$\text{GM2}_{\text{pars-inf}}$ : 2-state General Markov model, with only parsimony-informative characters observed

Parameters: Tree,  $2 \times 2$  Markov matrix on each edge,  
arbitrary root distribution

$\text{CFN}_{\text{pars-inf}}$ : Cavender-Farris-Neyman model, with only parsimony-informative characters observed

Submodel of  $\text{GM2}_{\text{pars-inf}}$  with symmetric Markov matrices,  
uniform root distribution

But much generalizes to  $k$ -state models,  $k > 2$  (in progress...)

## II: Identifiability:

For a fixed model,

Given an exact distribution of site-patterns arising from the model

— infinite amounts of 'perfect' data —

can we determine all model parameters?

Identifiability is necessary for **statistical consistency of inference**

Tree identifiability:

**Theorem** (Steel–Hendy–Penny, 1993): Identifiability of 4-taxon tree topologies fails for  $\text{CFN}_{\text{pars-inf}}$  (and hence for  $\text{GM2}_{\text{pars-inf}}$ ).

Proof is to explicitly give two parameter sets leading to same distribution of parsimony-informative patterns.

**Theorem** (Allman-Holder-R): Suppose all Markov matrix parameters are non-singular and have all positive entries. Then topologies of  $n$ -taxon trees are identifiable for  $\text{GM2}_{\text{pars-inf}}$  (and hence  $\text{CFN}_{\text{pars-inf}}$ ) for  $n \geq 8$ .

**Proof:**

- Enough to identify all 4-taxon subtrees.
- For subtree relating taxa  $a_1, a_2, a_3, a_4$ , fix some choice of parsimony-informative pattern at all *other* taxa
- Consider only patterns extending this choice to  $a_1, \dots, a_4$ .
- Observed frequencies of these extended patterns satisfy certain phylogenetic invariants depending on the 4-taxon topology.

(Invariants are inspired by the 4-point condition using a log-det distance – Cavender-Felsenstein, Steel)

*Note:* Identifiability of topologies for 5-, 6-, 7-taxon trees unknown.

## Numerical parameter identifiability:

Suppose

- the tree topology is known,
- all Markov matrix parameters are non-singular, and
- some parsimony-informative pattern has positive probability of being observed

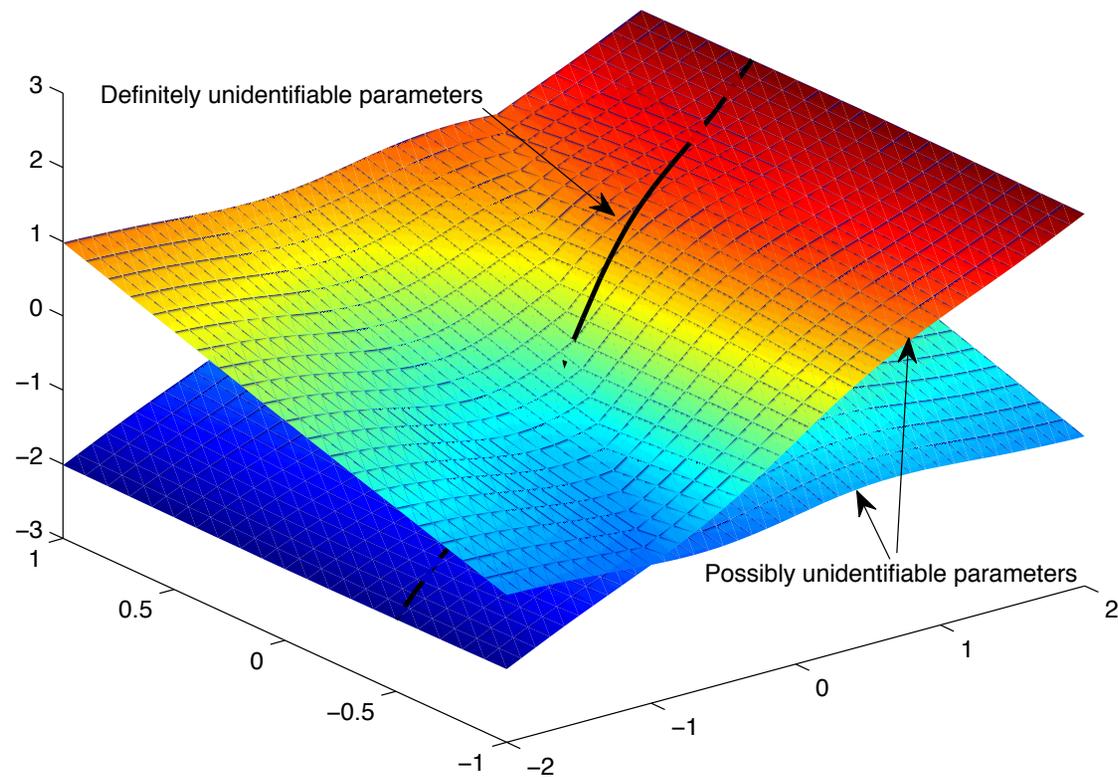
**Theorem** (Allman-Holder-R): For an  $n$ -taxon tree with  $n \geq 7$ , all numerical parameters of  $\text{GM2}_{\text{pars-inf}}$  are identifiable, up to ‘label-swapping’ at internal nodes. Hence numerical parameters of  $\text{CFN}_{\text{pars-inf}}$  are identifiable.

**Theorem** (Allman-Holder-R): For a 5-taxon tree **generic** numerical parameters of  $GM2_{\text{pars-inf}}$  are identifiable, up to 'label-swapping' at internal nodes.

However, there exists a subset of codimension 1 in the parameter space for which identifiability **may** fail.

Within this subset of potentially non-identifiable parameters, there is a smaller subset of codimension 2 in the full parameter space for which identifiability **definitely** fails.

# Cartoon of parameter space for 5-taxon trees:



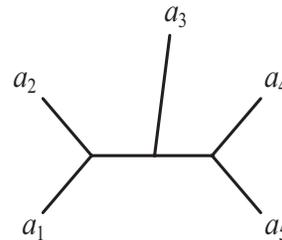
Specializing to  $\text{CFN}_{\text{pars-inf}}$ , generic parameters are identifiable.

However, the potentially non-identifiable parameters for 5-taxon trees include those from ultrametric (molecular clock) trees!

Sketch of method of proof of identifiability of numerical parameters:

We use

**Theorem** (Allman–R, 2008): For the 2-state General Markov model on a 5-taxon binary tree as shown, let  $\{0, 1\}$  denote the set of character states. Let  $p_{i_1 i_2 i_3 i_4 i_5}$  denote the joint probability of observing state  $i_j$  in the sequence at leaf  $a_j$ ,  $j = 1, \dots, 5$ .



Then the ideal of phylogenetic invariants for this model are generated by the  $3 \times 3$  minors of the following two matrices:

$$\begin{pmatrix} p_{00000} & p_{00001} & p_{00010} & p_{00011} & p_{00100} & p_{00101} & p_{00110} & p_{00111} \\ p_{01000} & p_{01001} & p_{01010} & p_{01011} & p_{01100} & p_{01101} & p_{01110} & p_{01111} \\ p_{10000} & p_{10001} & p_{10010} & p_{10011} & p_{10100} & p_{10101} & p_{10110} & p_{10111} \\ p_{11000} & p_{11001} & p_{11010} & p_{11011} & p_{11100} & p_{11101} & p_{11110} & p_{11111} \end{pmatrix}$$

and

$$\begin{pmatrix} P00000 & P00001 & P00010 & P00011 \\ P00100 & P00101 & P00110 & P00111 \\ P01000 & P01001 & P01010 & P01011 \\ P01100 & P01101 & P01110 & P01111 \\ P10000 & P10001 & P10010 & P10011 \\ P10100 & P10101 & P10110 & P10111 \\ P11000 & P11001 & P11010 & P11011 \\ P11100 & P11101 & P11110 & P11111 \end{pmatrix}.$$

If we have only probabilities  $q$  of patterns conditioned on parsimony-informativeness, then we know only *some* of these entries, but rescaled by an unknown factor.

$$\begin{pmatrix} \mathbf{q}_{00000} & \mathbf{q}_{00001} & \mathbf{q}_{00010} & q_{00011} & \mathbf{q}_{00100} & q_{00101} & q_{00110} & q_{00111} \\ \mathbf{q}_{01000} & q_{01001} & q_{01010} & q_{01011} & q_{01100} & q_{01101} & q_{01110} & \mathbf{q}_{01111} \\ \mathbf{q}_{10000} & q_{10001} & q_{10010} & q_{10011} & q_{10100} & q_{10101} & q_{10110} & \mathbf{q}_{10111} \\ q_{11000} & q_{11001} & q_{11010} & \mathbf{q}_{11011} & q_{11100} & \mathbf{q}_{11101} & \mathbf{q}_{11110} & \mathbf{q}_{11111} \end{pmatrix}$$

**Red** entries are unknown;  $3 \times 3$  minors must still be zero.

Judicious choices of  $3 \times 3$  minors allows for determination of unknown entries, provided certain  $2 \times 2$  minors don't vanish. E.g.,

$$\begin{vmatrix} q_{01001} & q_{01010} & q_{01011} \\ q_{10001} & q_{10010} & q_{10011} \\ q_{11001} & q_{11010} & \mathbf{q_{11011}} \end{vmatrix} = 0,$$

Expanding the determinant in cofactors by the last column we have

$$q_{01011} \begin{vmatrix} q_{10001} & q_{10010} \\ q_{11001} & q_{11010} \end{vmatrix} - q_{10011} \begin{vmatrix} q_{01001} & q_{01010} \\ q_{11001} & q_{11010} \end{vmatrix} + \mathbf{q_{11011}} \begin{vmatrix} q_{01001} & q_{01010} \\ q_{10001} & q_{10010} \end{vmatrix} = 0$$

Thus provided

$$\begin{vmatrix} q_{01001} & q_{01010} \\ q_{10001} & q_{10010} \end{vmatrix} \neq 0$$

we can determine  $\mathbf{q_{11011}}$  from other  $q_i$  where  $\mathbf{i} \in S$ .

For 5-taxon trees, enough  $2 \times 2$  minors may be zero to defeat this approach, but still gives understanding of potential non-identifiability.

For trees with at least 7 taxa, enough  $2 \times 2$  minors must be non-zero to determine all unknown entries.

Determining scaling factor is easy – sum of  $p_i$  is 1.