# **Reconstructing Phylogenetic Trees**

September 18th, 2008



### Systematic Methods Through Time





**Carolus Linneaus** 



**Charles** Darwin

![](_page_1_Picture_6.jpeg)

### Systematic Methods Through Time

![](_page_2_Figure_1.jpeg)

**Carolus Linneaus** 

**Charles** Darwin

![](_page_2_Picture_4.jpeg)

## **Computing Revolution**

- Late 1950s & 60s.
- Growing availability of core computing facilities.

![](_page_3_Picture_3.jpeg)

![](_page_3_Picture_4.jpeg)

### A QUANTITATIVE APPROACH TO A PROBLEM IN CLASSIFICATION <sup>1</sup>

CHARLES D. MICHENER AND ROBERT R. SOKAL<sup>2</sup> Department of Entomology, University of Kansas, Lawrence

Received October 10, 1956

![](_page_4_Picture_3.jpeg)

Charles D. Michener

![](_page_4_Picture_5.jpeg)

![](_page_4_Picture_6.jpeg)

Robert R. Sokal

![](_page_4_Picture_8.jpeg)

### A QUANTITATIVE APPROACH TO A PROBLEM IN CLASSIFICATION <sup>1</sup>

CHARLES D. MICHENER AND ROBERT R. SOKAL<sup>2</sup> Department of Entomology, University of Kansas, Lawrence

Received October 10, 1956

![](_page_5_Figure_3.jpeg)

Charles D. Michener

![](_page_5_Picture_6.jpeg)

### Tree Reconstruction I: Intro. & Distance Measures

- The challenge of tree reconstruction.
- Phenetics and an introduction to tree reconstruction methods.
  - Discrete v. distance measures.
  - Clustering v. optimality searches.
  - Tree building algorithms.

![](_page_6_Picture_6.jpeg)

# How Many Trees?

Taxa	Unrooted Trees	Rooted Trees
4	3	15
8	10,395	135,135
10	2,027,025	34,459,425
22	3x10^23	
50	3x10^74*	

\* More trees than there are atoms in the universe.

![](_page_7_Picture_3.jpeg)

## Reconstructing Trees

- The challenge of tree reconstruction.
  - Lots of possibilities.
- Phenetics and an introduction to tree reconstruction methods.
  - Discrete v. distance measures.
  - Clustering v. optimality searches.
  - Tree building algorithms.

![](_page_8_Picture_7.jpeg)

## **Discrete Data**

### discrete

sites

	1	2	3	4	5	6	7
1	Т	Т	A	Т	Т	A	A
2	A	А	Т	Т	Т	Α	Α
3	Α	А	А	Α	A	Т	А
4	A	Α	Α	Α	Α	Α	Т

sequences

![](_page_9_Picture_5.jpeg)

## **Discrete Data**

![](_page_10_Figure_1.jpeg)

![](_page_10_Picture_2.jpeg)

## Discrete v. Distance Trees

![](_page_11_Figure_1.jpeg)

![](_page_11_Picture_2.jpeg)

## **Clustering Methods**

![](_page_12_Figure_1.jpeg)

![](_page_12_Picture_2.jpeg)

# **Optimality Criterion**

![](_page_13_Figure_1.jpeg)

![](_page_13_Figure_2.jpeg)

![](_page_13_Picture_3.jpeg)

# **NP-Completeness**

- Non-deterministic polynomial.
- Impossible to guarantee optimal tree for even relatively modest number of sequences.
- Use of heuristic methods.

## **Available Methods**

![](_page_15_Figure_1.jpeg)

![](_page_15_Picture_2.jpeg)

### Distance Clustering Methods

- The phenetic approach.
- Two common algorithms for tree reconstruction.
  - UPGMA & Neighbor joining.

![](_page_16_Picture_4.jpeg)

### **Phenetics**

![](_page_17_Picture_1.jpeg)

- Also called **numerical taxonomy** because of emphasis on data.
- Relationships inferred based on overall similarity.

![](_page_17_Picture_4.jpeg)

### **Phenetics**

![](_page_18_Picture_1.jpeg)

![](_page_18_Picture_2.jpeg)

![](_page_19_Picture_0.jpeg)

- **UPGMA** Unweighted pair group method with arithmetic means (Sokal & Michener 1958).
- Remarkably simple and straightforward.
- Can be used with many types of distances (molecular, morphological, etc.).

![](_page_19_Picture_4.jpeg)

	A	В	С	D	E
А	0	32	48	51	50
В	32	0	26	34	29
С	48	26	0	42	44
D	51	34	42	0	44
E	50	29	44	44	0

![](_page_20_Picture_2.jpeg)

![](_page_21_Picture_0.jpeg)

	A	В	С	D	E
А	0	32	48	51	50
В	32	0	26	34	29
С	48	26	0	42	44
D	51	34	42	0	44
E	50	29	44	44	0

![](_page_21_Picture_2.jpeg)

	A	В	С	D	E
А	0	32	48	51	50
В	32	0	26	34	29
С	48	26	0	42	44
D	51	34	42	0	44
E	50	29	44	44	0

![](_page_22_Figure_2.jpeg)

![](_page_22_Picture_3.jpeg)

	A	В	С	D	E
А	0	32	48	51	50
В	32	0	26	34	29
С	48	26	0	42	44
D	51	34	42	0	44
E	50	29	44	44	0

A to 
$$B/C = (A \text{ to } B + A \text{ to } C)/2$$

$$40 = (32 + 48)/2$$

	▼					
	A	B/C	D	E		
А	0	40	51	50		
B/C	40	0	38	36.5		
D	51	38	0	44		
E	50	36.5	44	0		

![](_page_23_Picture_5.jpeg)

	А	В	С	D	E
А	0	32	48	51	50
В	32	0	26	34	29
С	48	26	0	42	44
D	51	34	42	0	44
E	50	29	44	44	0

A to 
$$B/C = (A \text{ to } B + A \text{ to } C)/2$$

C)/2 40 = (32 + 48)/2

	▼					
	A	B/C	D	E		
А	0	40	51	50		
B/C	40	0	38	36.5		
D	51	38	0	44		
E	50	36.5	44	0		

![](_page_24_Picture_5.jpeg)

#### **UPGMA Algorithm** 13 B 5.25 13 18.25 F. B/C Ε Α D 51 40 50 Α 0 B/C 38 36.5 40 0

![](_page_25_Picture_1.jpeg)

36.5

38

0

44

44

0

51

50

D

Ε

	A	B/C	D	E
А	0	40	51	50
B/C	40	0	38	36.5
D	51	38	0	44
E	50	36.5	44	0

![](_page_26_Picture_2.jpeg)

	А	B/C/E	D
А	0	45	51
B/C/E	45	0	37.25
D	51	37.25	0

![](_page_26_Picture_4.jpeg)

![](_page_27_Figure_1.jpeg)

	А	B/C/E	D
А	0	45	51
B/C/E	45	0	37.25
D	51	37.25	0

![](_page_27_Picture_3.jpeg)

	A	B/C/E	D
А	0	45	51
B/C/E	45	0	37.25
D	51	37.25	0

![](_page_28_Picture_2.jpeg)

	A	B/C/E/D
А	0	48
B/C/E/D	48	0

![](_page_28_Picture_4.jpeg)

![](_page_29_Figure_1.jpeg)

	A	B/C/E/D
А	0	48
B/C/E/D	48	0

![](_page_29_Picture_3.jpeg)

## **Distance Methods**

- Unweighted pair group means (Sokal & Michener 1958).
  - Easy to understand and implement even on large datasets.
  - Assumes molecular clock and prone to error if this is not the case.

	A	В	С	D	Ε
A	0	32	48	51	50
В	32	0	26	34	29
С	48	26	0	42	44
D	51	34	42	0	44
Е	50	29	44	44	0

![](_page_30_Figure_5.jpeg)

![](_page_30_Picture_6.jpeg)

![](_page_31_Figure_1.jpeg)

	A	В	С	D	E
А	0	15	21	16	11
В	15	0	10	9	8
С	21	10	0	15	14
D	16	9	15	0	9
E	11	8	14	9	0

![](_page_31_Picture_3.jpeg)

![](_page_32_Figure_1.jpeg)

	А	В	С	D	E
А	0	15	21	16	11
В	15	0	10	9	8
С	21	10	0	15	14
D	16	9	15	0	9
E	11	8	14	9	0

![](_page_32_Picture_3.jpeg)

![](_page_33_Picture_1.jpeg)

	A	В	С	D	E
А	0	15	21	16	11
В	15	0	10	9	8
С	21	10	0	15	14
D	16	9	15	0	9
E	11	8	14	9	0

![](_page_33_Picture_3.jpeg)

![](_page_34_Picture_1.jpeg)

	A	B/E	С	D
А	0	13	21	16
B/E	13	0	12	9
С	21	10	0	15
D	16	9	15	0

![](_page_34_Picture_3.jpeg)

![](_page_35_Figure_1.jpeg)

**Real Tree** 

**UPGMA** Tree

![](_page_35_Picture_4.jpeg)

![](_page_36_Figure_0.jpeg)

http://www.icp.be/~opperd/private/neighbor.html

![](_page_36_Picture_2.jpeg)

### Advantages of Distance Clustering Methods for Contemporary Systematics

- Computationally easy & quick.
- Typically give a single tree.
- Provide starting point for more sophisticated analyses.

![](_page_37_Picture_4.jpeg)

Disadvantages of Distance Clustering Methods & Phenetics for Contemporary Systematics

- Problems with distance measures.
- Problems with clustering algorithms.
- Problems with character choice.

![](_page_38_Picture_4.jpeg)

# Problems with Phenetics

![](_page_39_Picture_1.jpeg)

![](_page_39_Picture_2.jpeg)

![](_page_39_Picture_3.jpeg)

# Problems with Phenetics

![](_page_40_Picture_1.jpeg)

![](_page_40_Picture_2.jpeg)

![](_page_40_Picture_3.jpeg)

## **Types of Similarity**

### Homology v. Analogy

Similar due to inheritance Similar due to... uh...other factors

![](_page_41_Picture_4.jpeg)

http://evolution.berkeley.edu/evolibrary/

![](_page_41_Picture_6.jpeg)

## Homology

• Features shared due to inheritance from a common ancestor

![](_page_42_Picture_2.jpeg)

### **Vertebrate Forelimbs**

http://www.mun.ca/biology/scarr/

![](_page_42_Picture_5.jpeg)

## Analogies

## • Features shared due convergence or parallelism.

![](_page_43_Figure_2.jpeg)

### Swimming body form of vertebrates

http://evolution.berkeley.edu/evolibrary/

![](_page_43_Picture_5.jpeg)

![](_page_44_Figure_0.jpeg)

Mellorar T

![](_page_45_Figure_0.jpeg)

![](_page_45_Picture_2.jpeg)

# War of the Systematists

- Intense battles in the 1960s and 1970s.
- Considered in detail by the philosopher David Hull.

![](_page_46_Picture_3.jpeg)

SCIENCE as a PROCESS

![](_page_46_Picture_5.jpeg)

DAVID L. HULL

![](_page_46_Picture_7.jpeg)

![](_page_46_Picture_8.jpeg)

![](_page_46_Picture_10.jpeg)

### **The Cladistic Revolution**

• **Cladistics:** Hierarchical classification of species based on evolutionary ancestry.

![](_page_47_Picture_2.jpeg)

### **The Cladistic Revolution**

- Willi Hennig the founder of cladistics.
  - <u>Phylogenetic Systematics</u>
    - 1950 (German edition).
    - 1966 (English translation).

![](_page_48_Picture_5.jpeg)

Willi Hennig

![](_page_48_Picture_7.jpeg)

# The Cladogram

• **Cladogram:** A branching diagram that depicts a set of hypothesized evolutionary relationships (i.e., a phylogeny).

![](_page_49_Figure_2.jpeg)

![](_page_49_Picture_3.jpeg)

## **Clades & Cladists**

• **Clade:** A monophyletic group.

![](_page_50_Figure_2.jpeg)

©1999 Addison Wesley Longman, Inc.

![](_page_50_Picture_4.jpeg)

## **Clades & Cladists**

• **Cladist:** One who uses cladistics to infer phylogenetic relationships.

![](_page_51_Picture_2.jpeg)

Arnold Kluge

![](_page_51_Picture_4.jpeg)

![](_page_51_Picture_5.jpeg)

Joel Cracraft

![](_page_51_Picture_7.jpeg)

![](_page_51_Picture_8.jpeg)

Steve Poe

![](_page_51_Picture_10.jpeg)

Kevin De Queiroz

![](_page_51_Picture_12.jpeg)

Norman Platnick

![](_page_51_Picture_14.jpeg)

# **Cladistic Principles**

- I. Use characters with special evolutionary significance.
- II. Reconstruction of phylogeny based on most parsimonious reconstruction of these characters.

![](_page_52_Figure_3.jpeg)

53

![](_page_52_Picture_4.jpeg)

## Cladistic Terms for Characters

![](_page_53_Figure_1.jpeg)

- Instead of being 'primitive' or 'advanced', characters are 'plesiomorphic' and 'apomorphic.'
- **Apomophy:** A derived character.
- **Plesiomorphy:** An ancestral or primitive character.

![](_page_53_Picture_5.jpeg)

## Cladistic Terms for Characters

![](_page_54_Figure_1.jpeg)

• **Synapomorphy:** An apomorphy shared among taxa due to common ancestry.

## Cladistic Terms for Characters

![](_page_55_Picture_1.jpeg)

- **Synapomorphy:** An apomorphy shared among taxa due to common ancestry.
- **Symplesiomorphy:** A plesiomorphy shared among taxa, but which predates their common ancestor.

![](_page_55_Picture_4.jpeg)

### Classification with Symplesiomorphy

![](_page_56_Figure_1.jpeg)

![](_page_56_Picture_2.jpeg)

### **Cladistic Terms**

#### Correct: grouped by synapomorphy

![](_page_57_Figure_2.jpeg)

Incorrect: grouped by symplesiomorphy

![](_page_57_Figure_4.jpeg)

![](_page_57_Picture_5.jpeg)

## **Cladistic Terms**

![](_page_58_Figure_1.jpeg)

• Autapomorphy: A derived trait that is unique to one group.

![](_page_58_Picture_3.jpeg)

## Hierarchy of <sup>60</sup> Cladistic Characters

clade c: vascular plants

![](_page_59_Figure_1.jpeg)

http://www.ucmp.berkeley.edu/IB181/VPL/Phylo/Phylo2.html

![](_page_59_Picture_3.jpeg)

### Identifying Synapomorphies

- Similar fundamental structure.
- Same relation to surrounding characters.
- Similar development.
- Product of natural selection?

![](_page_60_Picture_5.jpeg)

### Methods

![](_page_61_Figure_1.jpeg)

![](_page_61_Picture_2.jpeg)

### Methods

![](_page_62_Figure_1.jpeg)

![](_page_62_Picture_2.jpeg)

## Parsimony

- Preference for the tree with the "minimum net amount of evolution" (Edwards and Cavalli-Sforza 1963)
- Seen as an extension of Occam's razor.
  - entia non sunt multiplicanda praeter necessitatem
  - "entities should not be multiplied beyond necessity" or "all things being equal, the simplest solution tends to be the best one."

![](_page_63_Picture_5.jpeg)

![](_page_63_Picture_6.jpeg)

## Parsimony Reconstruction

	1	2	3	4	5	6
Alpha	1	0	0	1	1	0
Beta	0	0	1	0	0	0
Gamma	1	1	0	0	0	0
Delta	1	1	0	1	1	1
Eps	0	0	1	1	1	0

![](_page_64_Picture_2.jpeg)

## Parsimony Reconstruction

	1	2	3	4	5	6
Alpha	1	0	0	1	1	0
Beta	0	0	1	0	0	0
Gamma	1	1	0	0	0	0
Delta	1	1	0	1	1	1
Eps	0	0	1	1	1	0

![](_page_65_Figure_2.jpeg)

![](_page_65_Picture_3.jpeg)

## Parsimony Reconstruction

	1	2	3	4	5	6
Alpha	1	0	0	1	1	0
Beta	0	0	1	0	0	0
Gamma	1	1	0	0	0	0
Delta	1	1	0	1	1	1
Eps	0	0	1	1	1	0

![](_page_66_Figure_2.jpeg)

![](_page_67_Figure_0.jpeg)

![](_page_67_Picture_1.jpeg)

Hillis et al. 1994

## **Cladistics Today**

- The Willi Hennig Society.
- The journal Cladistics.
- Cladisitcs and Cladists with a capital 'C.'

![](_page_68_Picture_4.jpeg)

![](_page_68_Picture_5.jpeg)