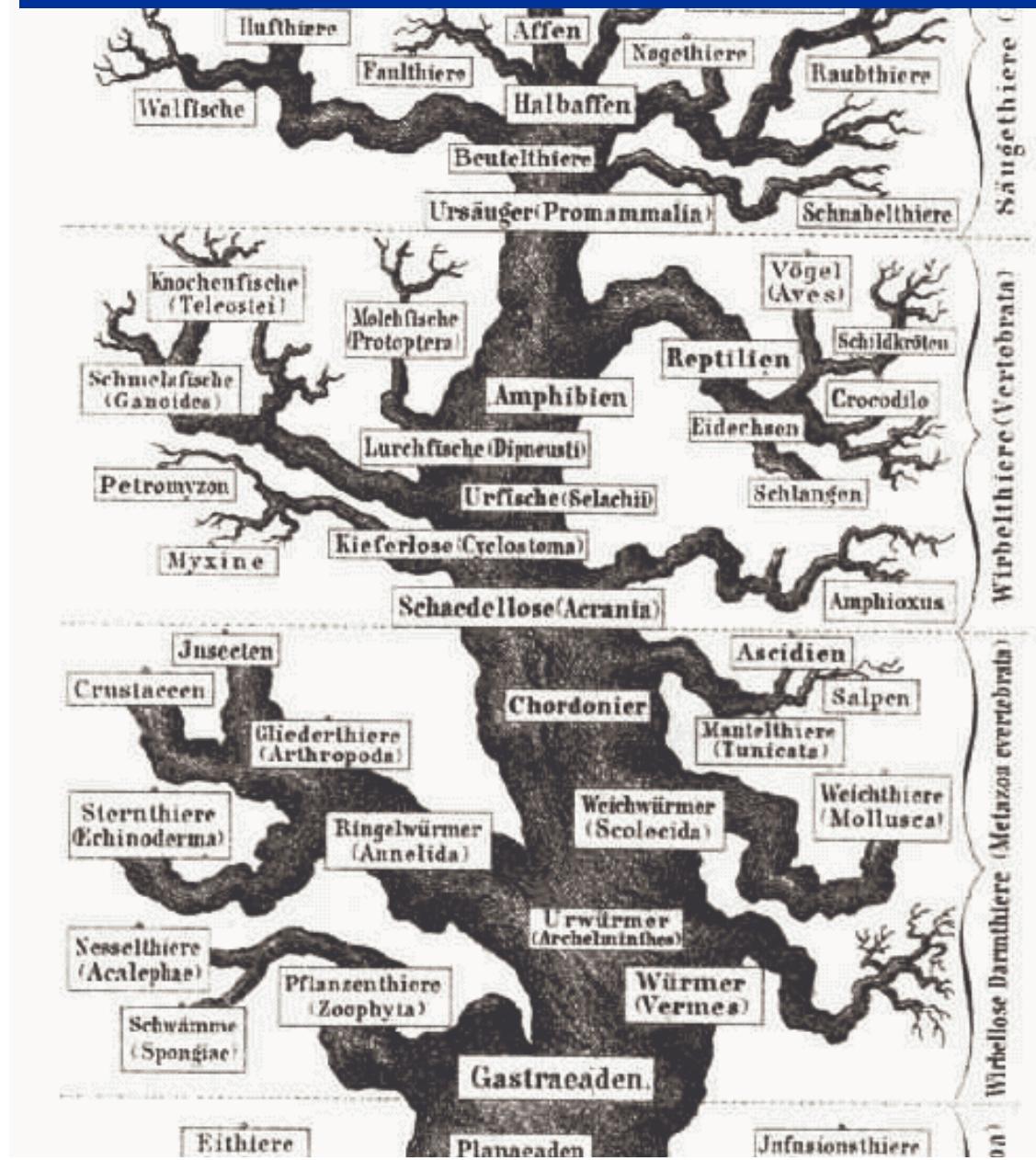
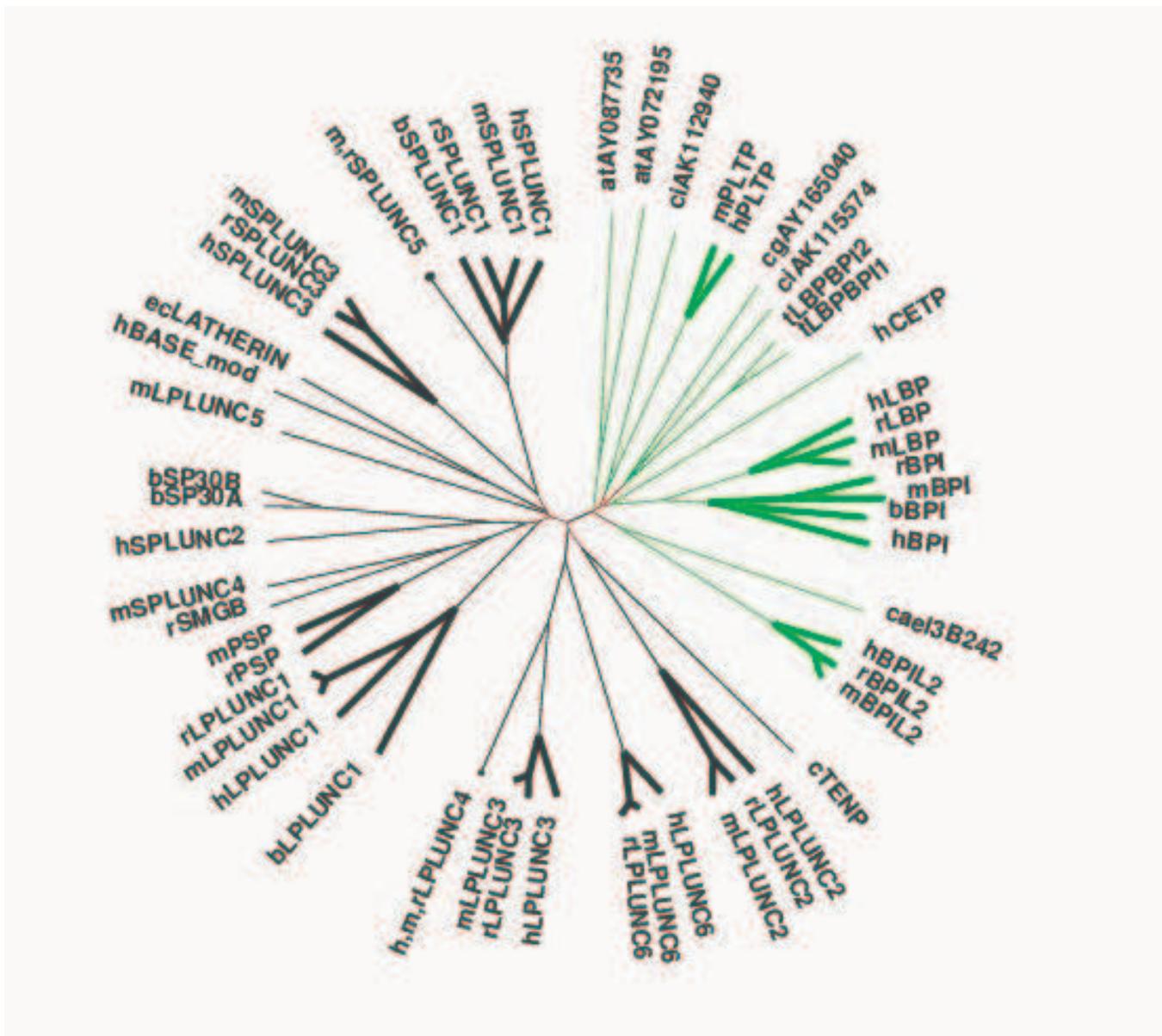


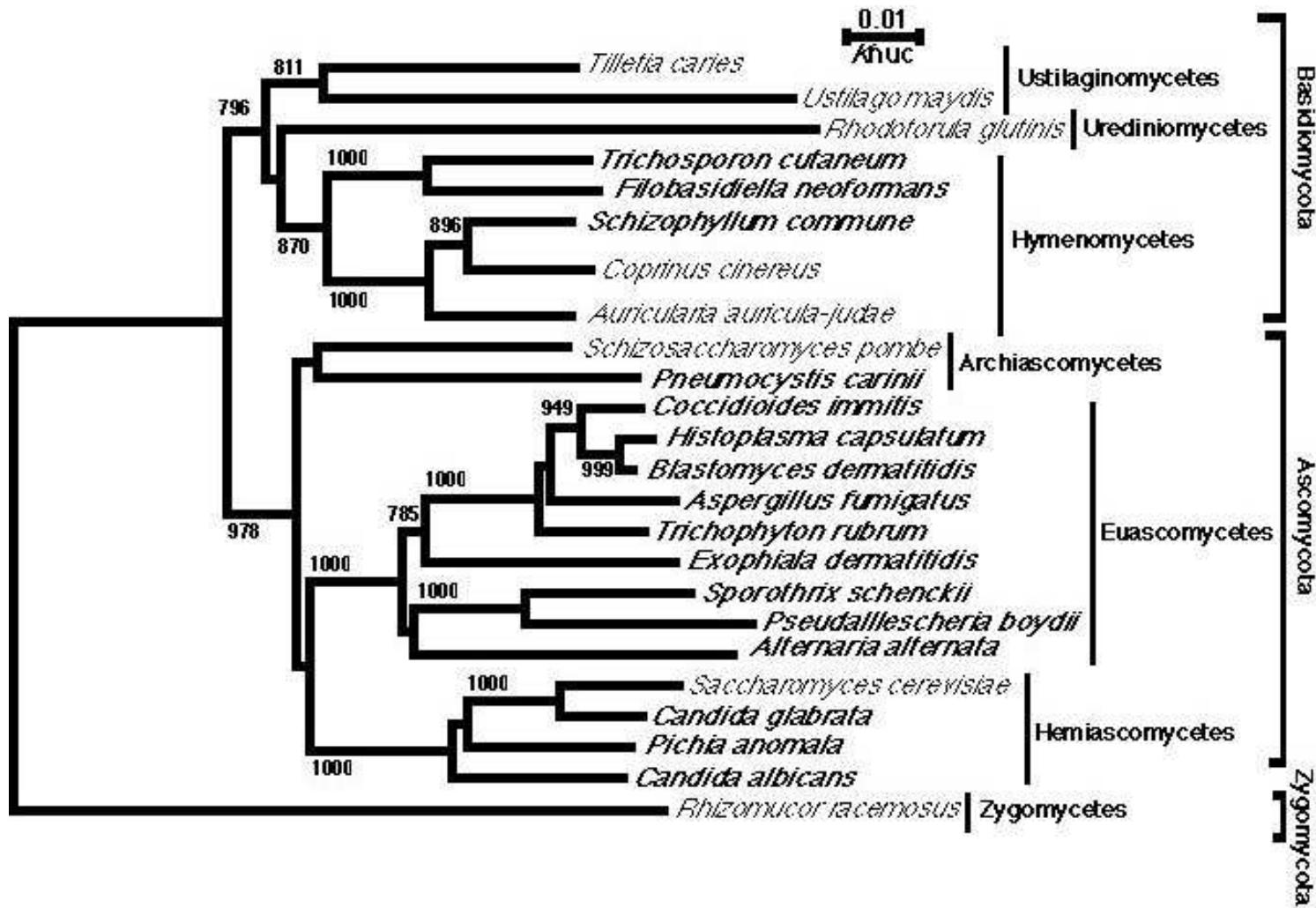
# phylogenetic trees - morphological



# phylogenetic trees - genetic



# phylogenetic trees - genetic



# Phylogeny

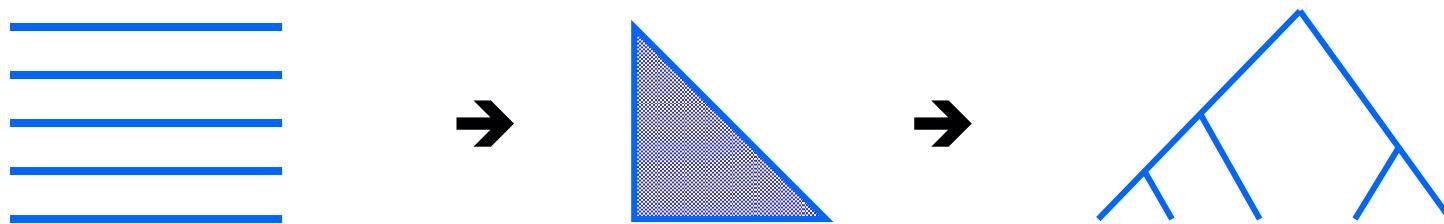


Charlie Chaplin once entered a Charlie Chaplin look-alike contest in Monte Carlo, Monaco. He placed third.

# phylogeny

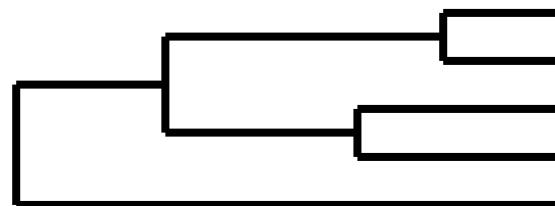
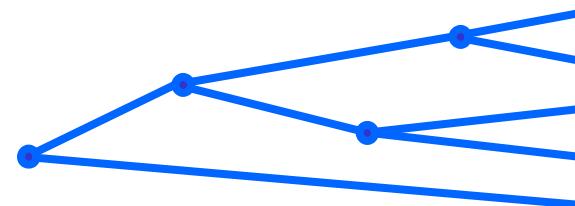
## general approach

- > choose sequences (taxa)
- > multiple alignment
  - ⇒ character matches
  - ⇒ distances
- > find tree (topology + edge lengths)

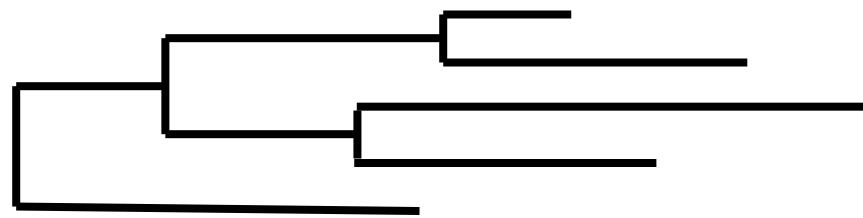
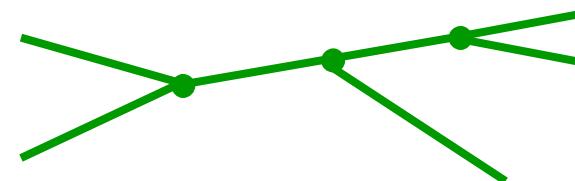


# trees

- > leaves: taxa '*operational taxonomic units*'  
internal: hypothetical
- > binary bifurcation, independent visual representation



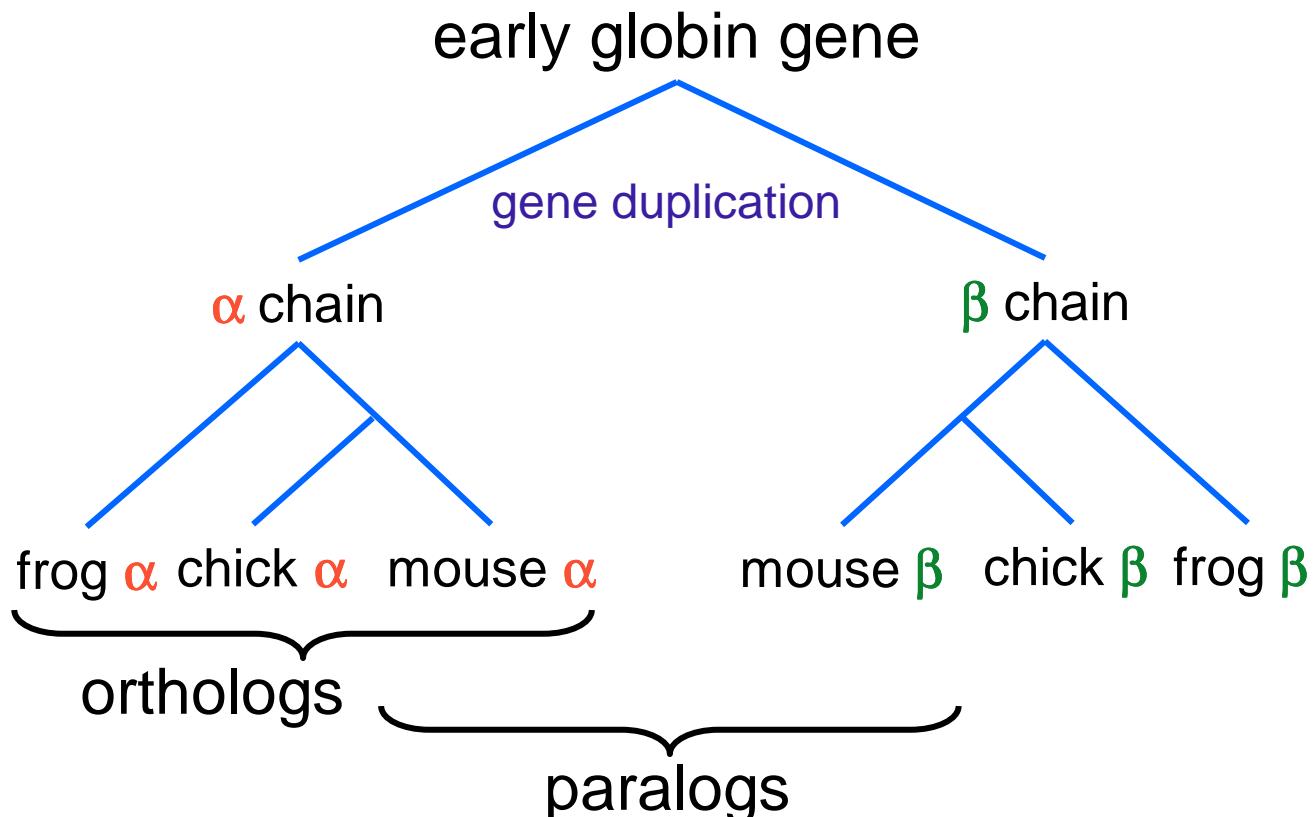
- > rooted vs. unrooted



- > branch length (edges) ~ evolutionary model / clock
- > too many trees !!  
assumptions on tree or heuristics

# homologous genes

homologous genes ~ corresponding  
orthologous in different species  
paralogous in same species (genome)  $\alpha \beta \gamma \delta \varepsilon \dots$





## Phylogeny Programs

table of contents by methods available:

General-purpose • Parsimony • Distance matrix • Computation of distances • Maximum likelihood, Bayesian • Quartets • Genetic algorithms • Evolutionary Parsimony • Interactive tree manipulation • Looking for hybridization • Bootstrapping • Compatibility analysis • Consensus trees, distances • Tree-based alignment • Gene duplication • Biogeographic analysis • Comparative method analysis • Simulation of trees • Examination of shapes • Clocks, dating, stratigraphy • Prediction of data from trees • Tree plotting/drawing • Teaching • Web or e-mail servers

<http://evolution.genetics.washington.edu/phylip/software.html>

# phylogeny

chimpanzee	AAGCTTCACC	GGCGCAATTA	TCCTCATAAT	CGCCCACGGA	CTTACACATCCT
gibbon	T	A	T	CCG	
gorillae			G	T	T
homosapien		G C	T		
orangutang	CC	C		G	T

Neighbor-joining method

-----gorillae

--3homosapien

|  
|-----chimpanzee

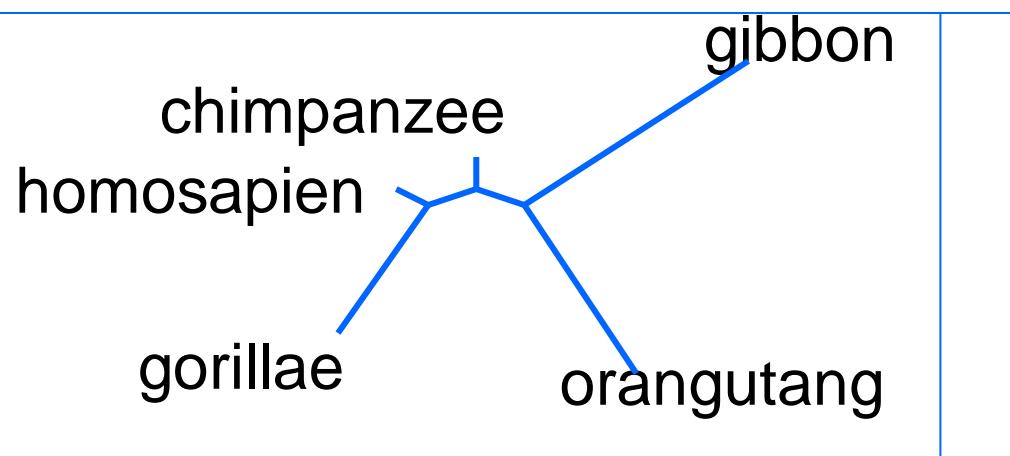
|-2

|  
|-----gibbon

|-1

|  
|-----orangutang

chimpanzee					
gibbon	0.21				
gorillae	0.13	0.29			
homosapien	0.06	0.24	0.11		
orangutang	0.15	0.26	0.25	0.18	



remember: this is an unrooted tree!

# contents

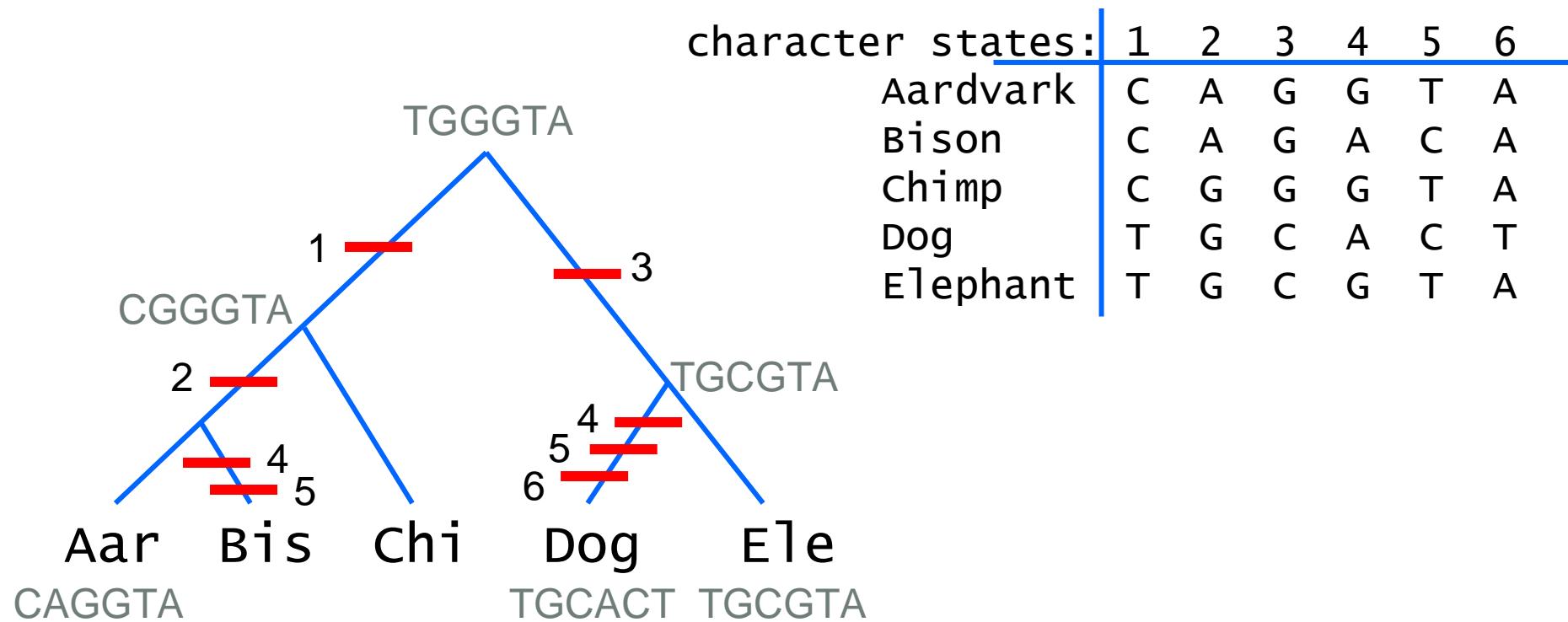
## this lecture

- Character based
  - ❖ Sankoff
  - ❖ Maximum Likelihood Method
  - ❖ Heuristic Approaches
- Distance based
  - ❖ Neighbour Joining
  - ❖ UPGMA
  - ❖ Heuristic Approaches

## I

## character based

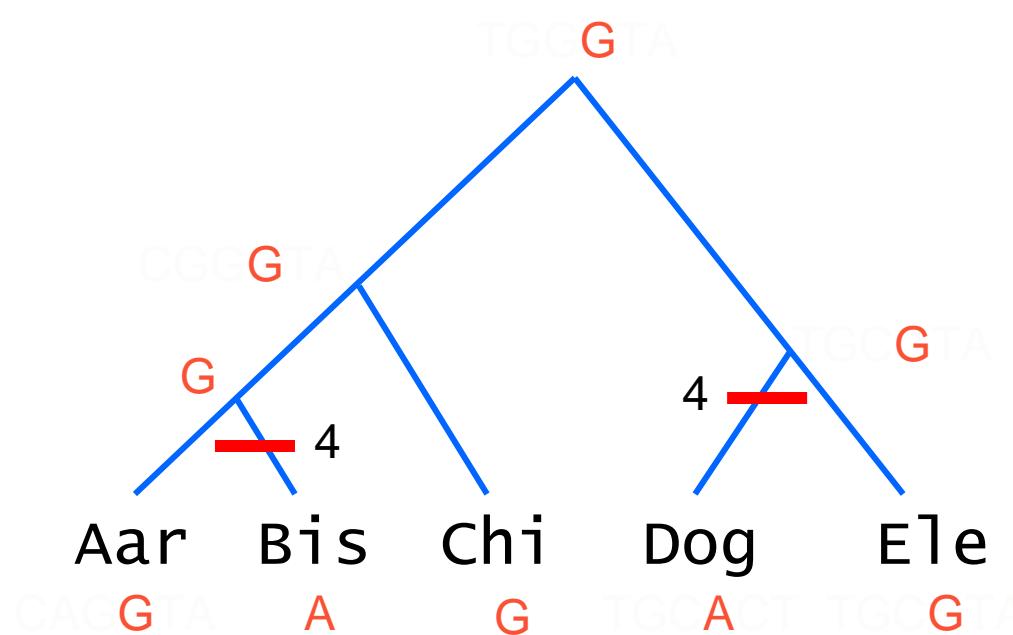
**parsimony:** minimal number of changes  
5 species, 6 characters



**small** parsimony: topology given, find labeling  
**large** parsimony: also find optimal topology

# character based

small parsimony: topology given  
find labeling for character 4



character states:

	1	2	3	4	5	6
Aardvark	C	A	G	G	T	A
Bison	C	A	G	A	C	A
Chimp	C	G	G	G	T	A
Dog	T	G	C	A	C	T
Elephant	T	G	C	G	T	A

small parsimony (single character)

given topology: rooted tree

- k species - k states [in leaves]
- cost for changing state  $C_{ij}$

bottom-up tree evaluation:

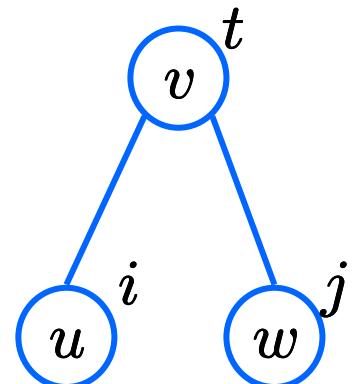
assign a cost vector  $S_t(v)$  at each node  
= minimal cost of state  $t$  at node  $v$

✓ leaf

$$S_t(v) = \begin{cases} 0 & \text{state}(v) = t \\ \infty & \text{otherwise} \end{cases}$$

✓ internal

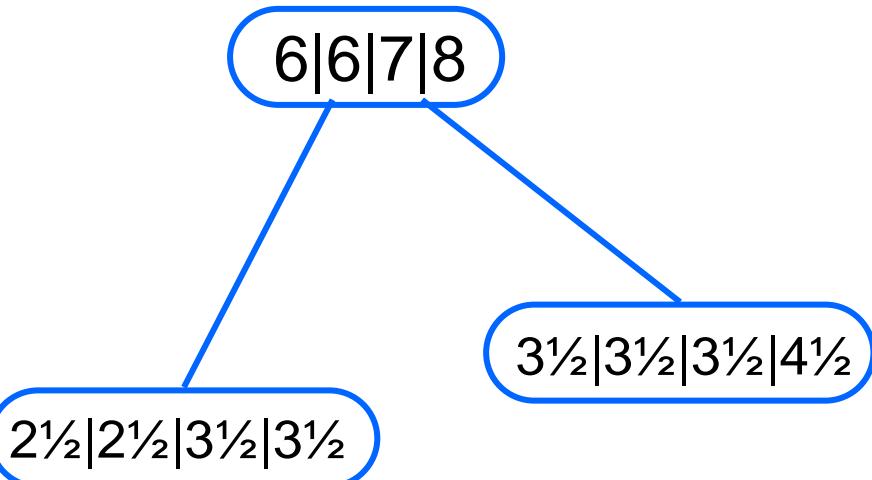
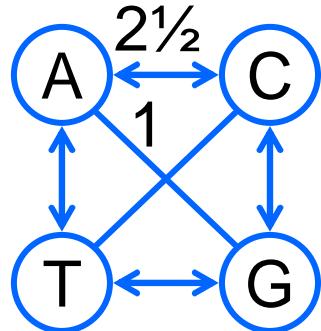
$$S_t(v) = \min_i \{C_{ti} + S_i(u)\} + \min_j \{C_{tj} + S_j(w)\}$$



to be repeated for each character

# character based

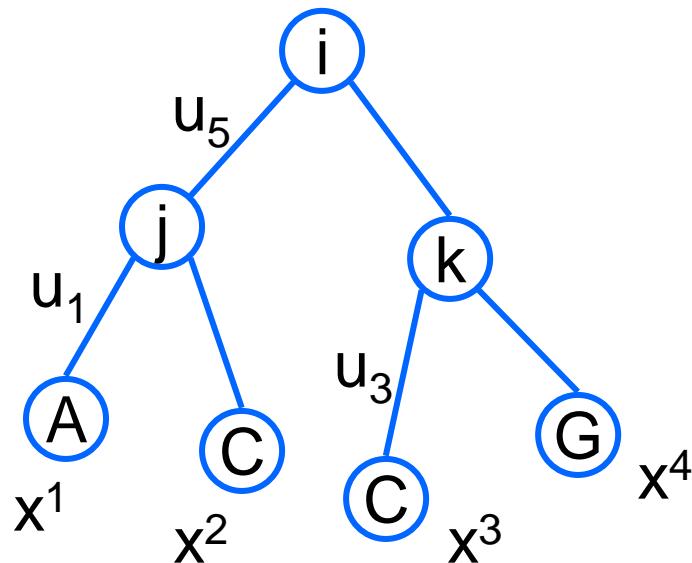
# Sankoff



$A|C|G|T$   
 $S_t(v)$

$$\begin{aligned}
 A & 6 \quad \min \{ 2\frac{1}{2}+0, 2\frac{1}{2}+2\frac{1}{2}, 3\frac{1}{2}+1, 3\frac{1}{2}+2\frac{1}{2} \} \\
 & + \min \{ 3\frac{1}{2}+0, 3\frac{1}{2}+2\frac{1}{2}, 3\frac{1}{2}+1, 4\frac{1}{2}+2\frac{1}{2} \} \\
 C & 6 \quad \min \{ 2\frac{1}{2}+2\frac{1}{2}, 2\frac{1}{2}+0, 3\frac{1}{2}+2\frac{1}{2}, 3\frac{1}{2}+1 \} \\
 & + \min \{ 3\frac{1}{2}+2\frac{1}{2}, 3\frac{1}{2}+0, 3\frac{1}{2}+2\frac{1}{2}, 4\frac{1}{2}+1 \} \\
 G & 7 \quad \min \{ 2\frac{1}{2}+1, 2\frac{1}{2}+2\frac{1}{2}, 3\frac{1}{2}+0, 3\frac{1}{2}+2\frac{1}{2} \} \\
 & + \min \{ 3\frac{1}{2}+1, 3\frac{1}{2}+2\frac{1}{2}, 3\frac{1}{2}+0, 4\frac{1}{2}+2\frac{1}{2} \} \\
 T & 8 \quad \min \{ 2\frac{1}{2}+2\frac{1}{2}, 2\frac{1}{2}+1, 3\frac{1}{2}+2\frac{1}{2}, 3\frac{1}{2}+0 \} \\
 & + \min \{ 3\frac{1}{2}+2\frac{1}{2}, 3\frac{1}{2}+1, 3\frac{1}{2}+2\frac{1}{2}, 4\frac{1}{2}+0 \}
 \end{aligned}$$

# maximum likelihood



A	A	G	x <sup>1</sup>
C	T	C	x <sup>2</sup>
C	G	G	x <sup>3</sup>
G	G	A	x <sup>4</sup>

four taxa, three sites

site specific probability, multiply over all sites  
probabilities  $p_{ij}(t)$  given by **evolutionary model**

$$P(D|T) = \sum_{ijk} p_{ij}(u_5) \dots p_{kC}(u_3) \dots$$

# evolutionary models

## continuous time Markov model

### Jules-Cantor

instantaneous change

$$q_{xx} = -\frac{3}{4}\alpha \quad q_{xy} = \frac{1}{4}\alpha \quad \text{evolutionary rate } \alpha$$

- $p_{xx}(t) = \frac{1}{4} + \frac{3}{4} \exp(-t\alpha)$
- $p_{xy}(t) = \frac{1}{4} - \frac{1}{4} \exp(-t\alpha)$

### Kimura

transitions A – C – G – T – A  
transversions A – G, T – C

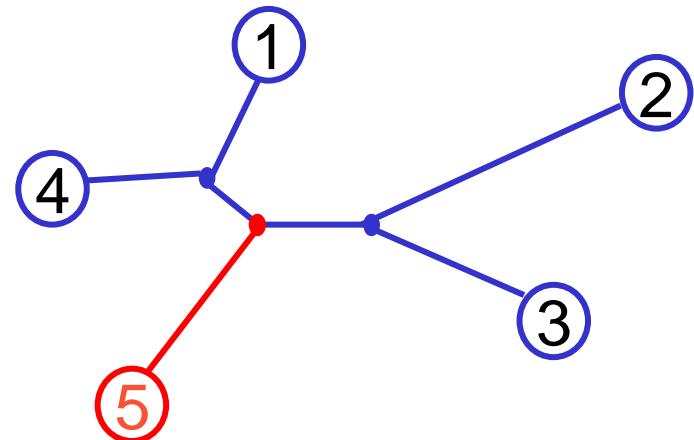
### Felsenstein, HKY, ...

# character based

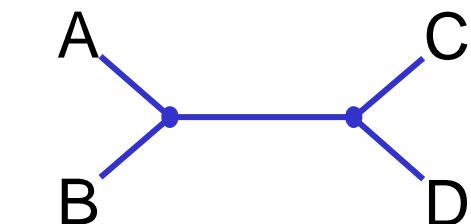
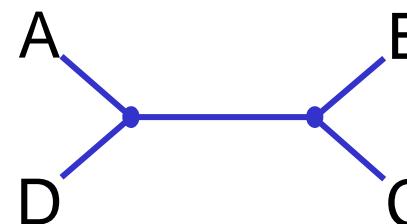
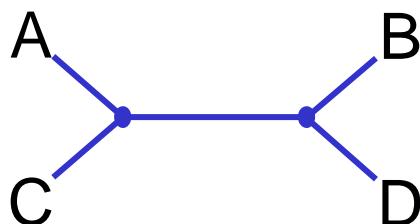
# heuristics

large parsimony is NP complete 😞  
search strategies, heuristics  
space = 'all' trees

→ branch-and-bound  
add taxon (species) at a time



→ nearest neighbour interchanges



metric space objects + distance  $d$

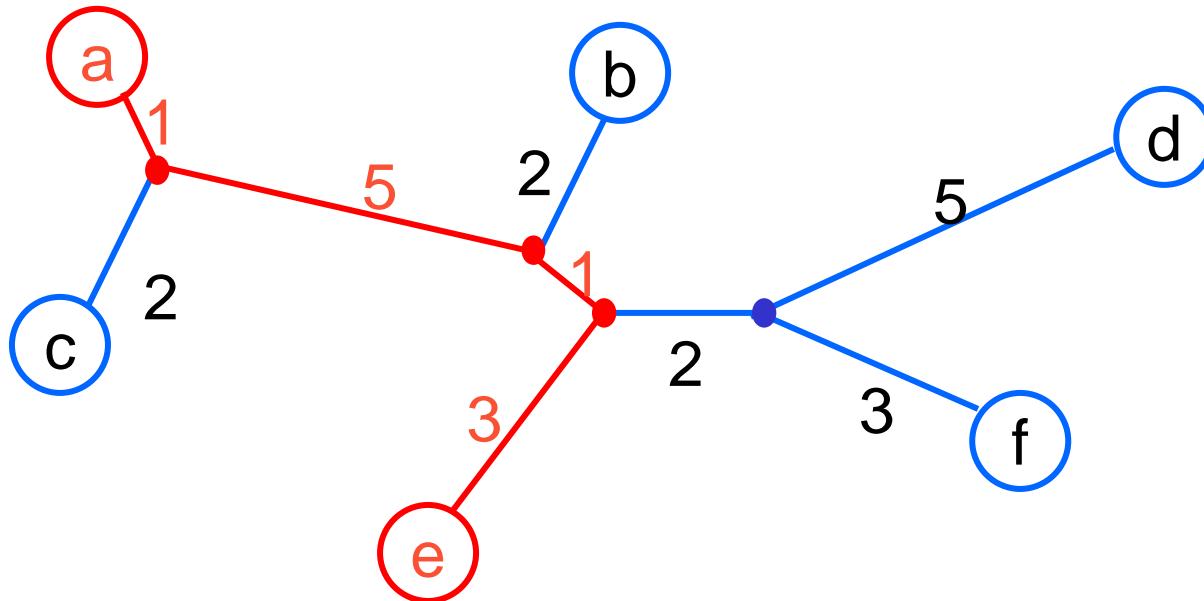
- $d(x,y) \geq 0$
  - $d(x,y) = 0$  iff  $x=y$
  - $d(x,y) = d(y,x)$  symmetric
- triangle inequality
- $d(x,z) \leq d(x,y) + d(y,z)$

additional properties  $\Rightarrow$  specific algorithms

- triangle inequality
- additive (Neighbour Joining)
- ultrametric (UPGMA)

distance based

additive



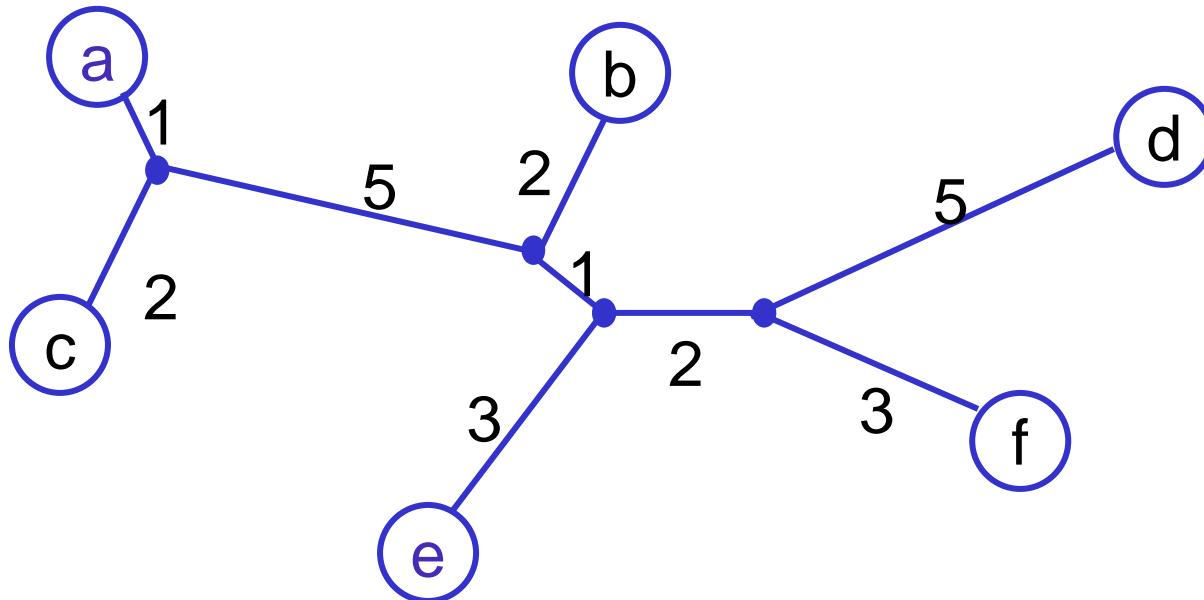
additive distance function = generated by a tree

$$d(a,e) = 1+5+1+3=10$$

- given a distance matrix  
how to recognize it is additive ?
- given an additive distance matrix  
how to (re)construct the tree ?

distance based

additive



	a	b	c	d	e	f
a	-	8	3	14	10	12
b		-	9	10	6	8
c			-	15	11	13
d				-	10	8
e					-	8
f						-

# distance based

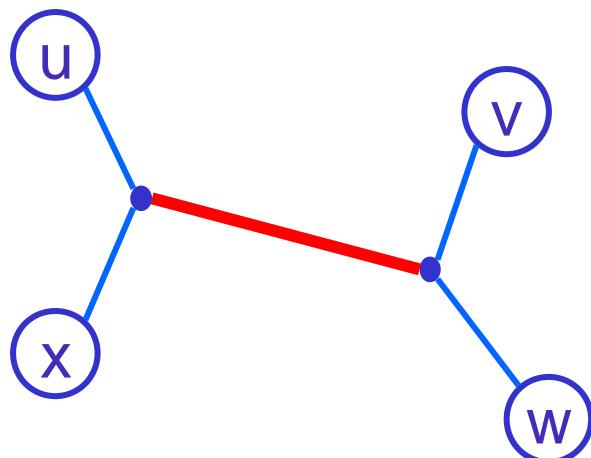
# additive

- given a distance matrix  
how to recognize it is additive ?

four point condition  $\Leftrightarrow$  additive

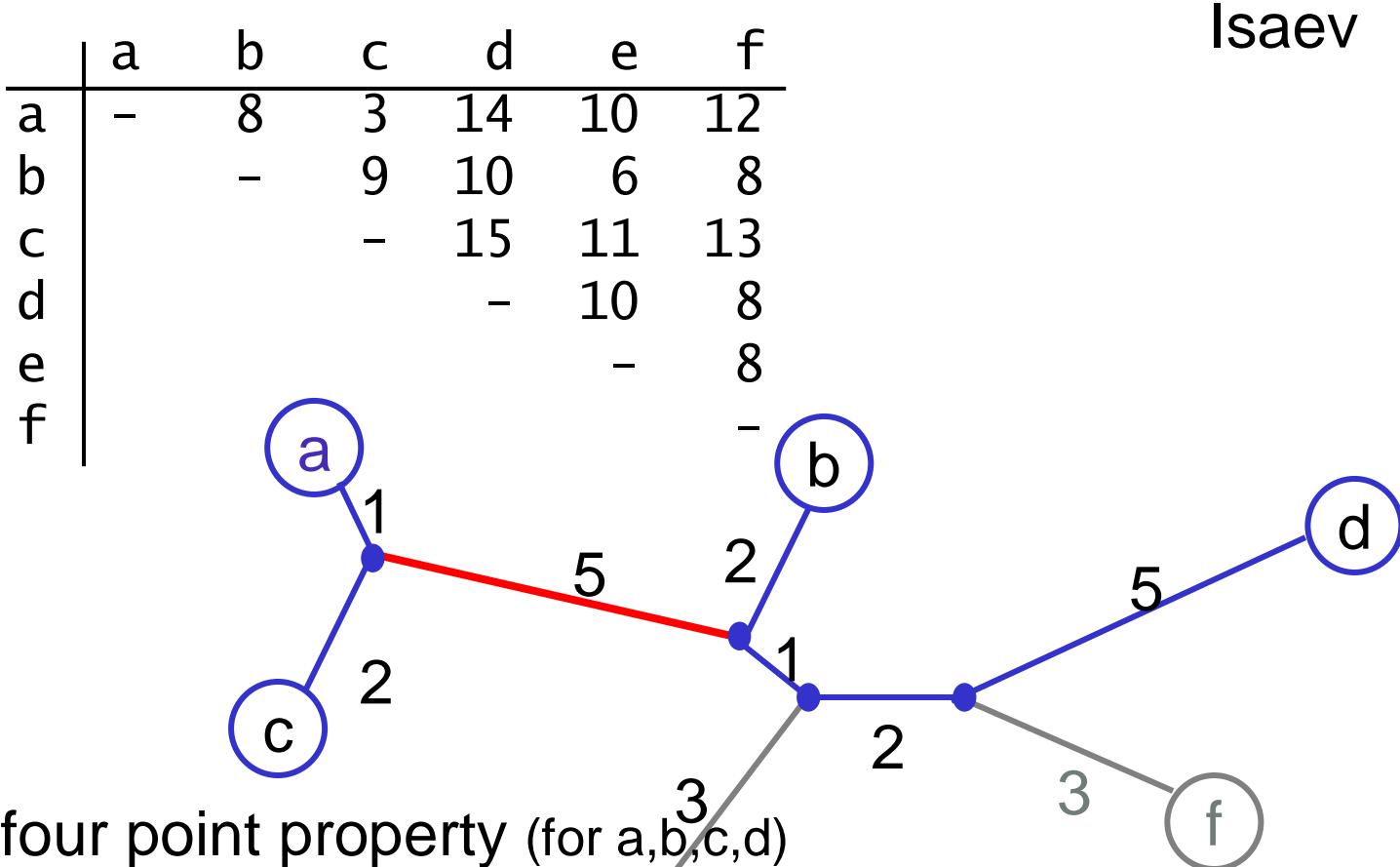
$$d(u,v) + d(w,x) = d(u,w) + d(v,x) \geq d(u,x) + d(v,w)$$

for some ordering of  $u,v,w,x$



# distance based

# additive



$$d(a,b) + d(c,d) = 8 + 15 = 23$$

$$d(a,c) + d(b,d) = 3 + 10 = 13$$

$$d(a,d) + d(b,c) = 14 + 9 = 23$$

## distance based

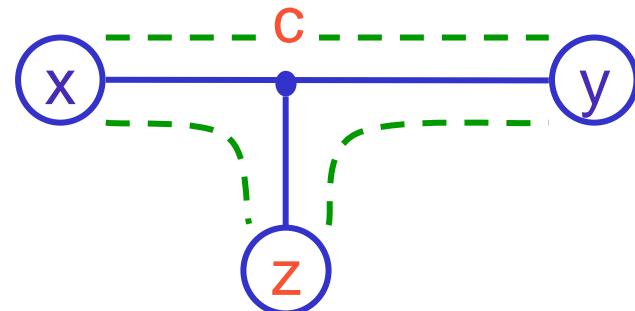
- given an additive distance matrix  
how to (re)construct the tree ?

intuition: adding the third point ... where?

$$d(x, y) = \ell_{xc} + \ell_{cy}$$

$$d(x, z) = \ell_{xc} + \ell_{cz}$$

$$d(y, z) = \ell_{yc} + \ell_{cz}$$



$$\ell_{cy} = \frac{d(x, y) + d(y, z) - d(x, z)}{2}$$

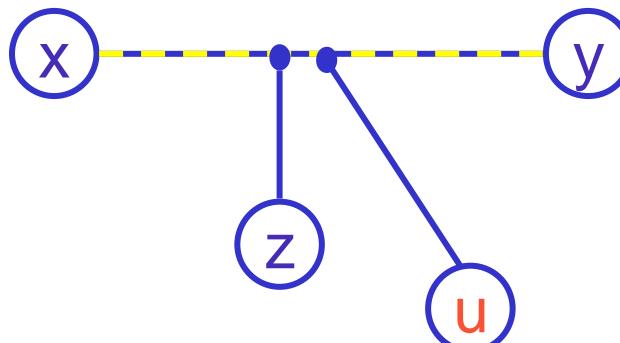
$\geq 0$  triangle inequality

and symmetric for  $cx, cz$

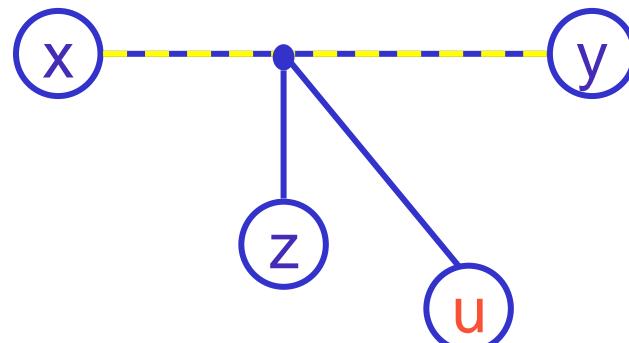
## distance based

- given an additive distance matrix  
how to (re)construct the tree ?

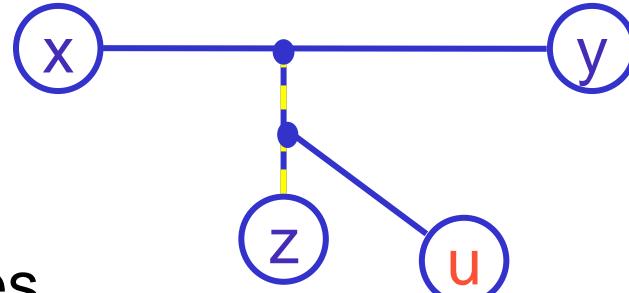
adding **another** point to (x,y)



ok



try again for (x,z)

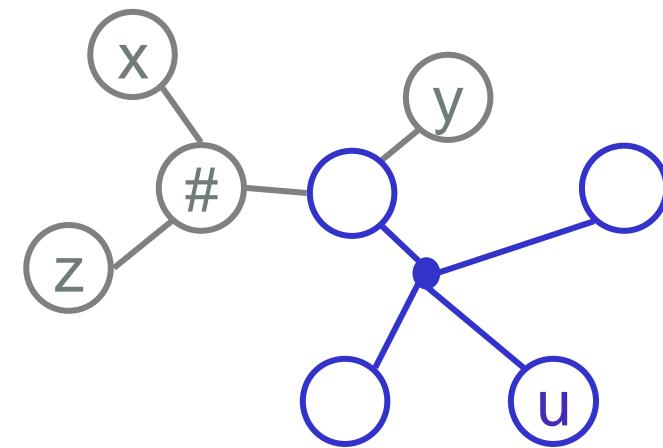
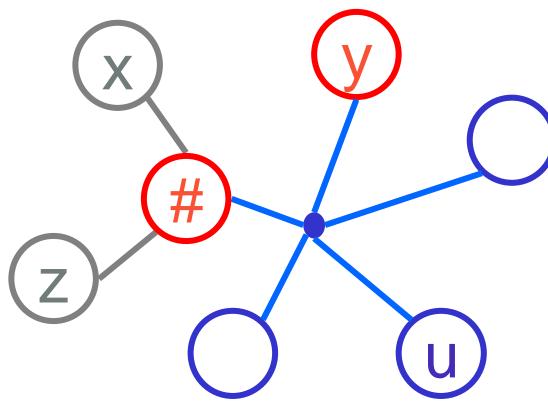
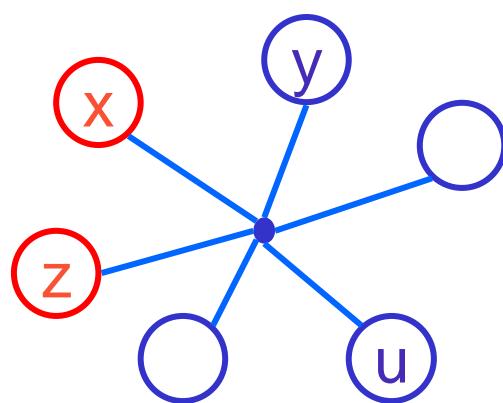


in reality errors in distances ...

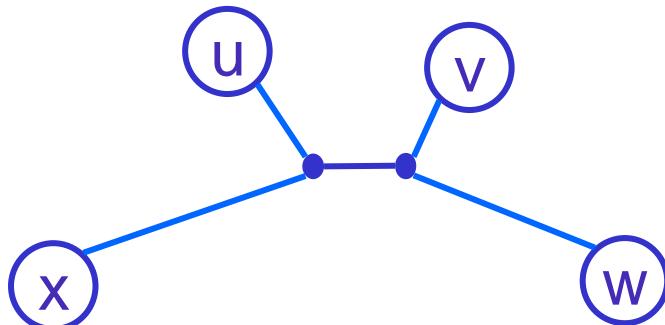
# distance based

# neighbour joining

intuition: star graph, join closest neighbours



shortest distance  $\Rightarrow$  neighbours



## distance based

## neighbour joining

total distance to other nodes

pairs of nodes

$$r(x) = \frac{1}{N-2} \sum_z d(x, z)$$

$$D(x, y) = r(x) + r(y) - d(x, y)$$

- > pick  $x, y$  maximal for  $D(x, y)$
- > join  $x, y$  into a new node  $z$

distances of  $z$  to 'old' nodes:

$$d(z, x) = \frac{1}{2}( d(x, y) + r(x) - r(y) )$$

$$d(z, y) = \dots$$

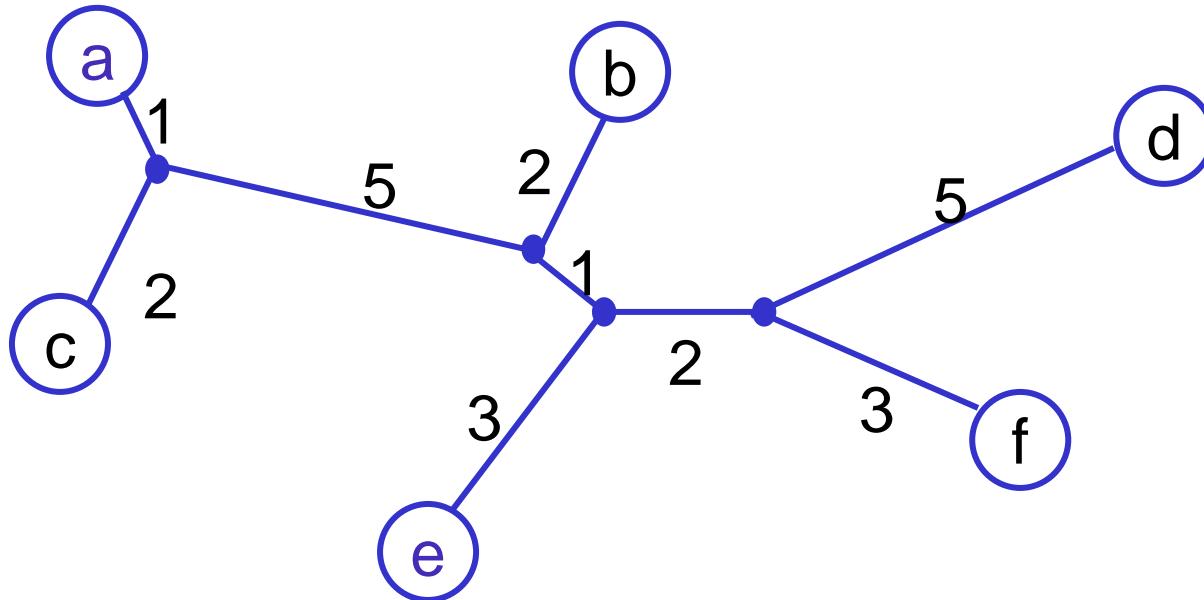
$$\text{thus } d(x, z) + d(z, y) = d(x, y)$$

$$d(z, w) = \frac{1}{2}( d(x, w) + d(y, w) - d(x, y) )$$

$$\text{thus } \geq 0$$

# neighbour joining

example

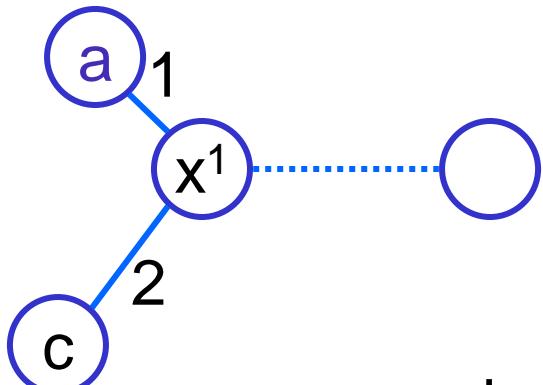


	a	b	c	d	e	f
a	-	8	3	14	10	12
b		-	9	10	6	8
c			-	15	11	13
d				-	10	8
e					-	8
f						-

# neighbour joining

example

	a	b	c	d	e	f	$4 \cdot r$	$N-2=4$
a	-	8	3	14	10	12	47	
b		-	9	10	6	8	41	
c			-	15	11	13	51	
d				-	10	8	57	
e					-	8	45	
f						-	49	



$4 \cdot D(x,y)$

	a	b	c	d	e	f
a	-	56	86	48	52	48
b		-	56	58	32	58
c			-	48	52	48
d				-	62	74
e					-	62
f						-

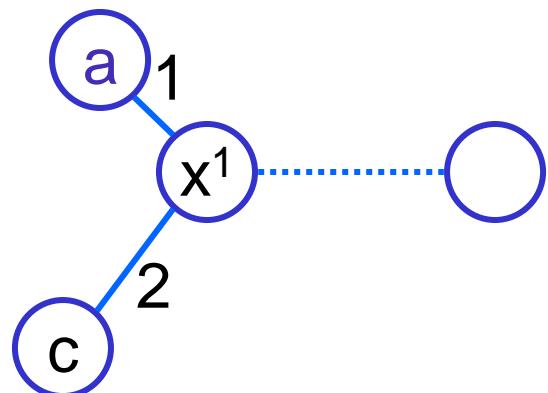
new node  $x^1 = \{a,c\}$

$$d(a,x^1) = \frac{1}{2} ( 3 + \frac{47}{4} - \frac{51}{4} ) = 1$$

$$d(c,x^1) = \frac{1}{2} ( 3 + \frac{51}{4} - \frac{47}{4} ) = 2$$

# neighbour joining

# example



new node  $x^1 = \{a,c\}$

$$d(x^1, b) = \frac{1}{2}(8 + 9 - 3) = 7$$

$$d(x^1, d) = \frac{1}{2}(14+15-3) = 13$$

distance based

ultrametric

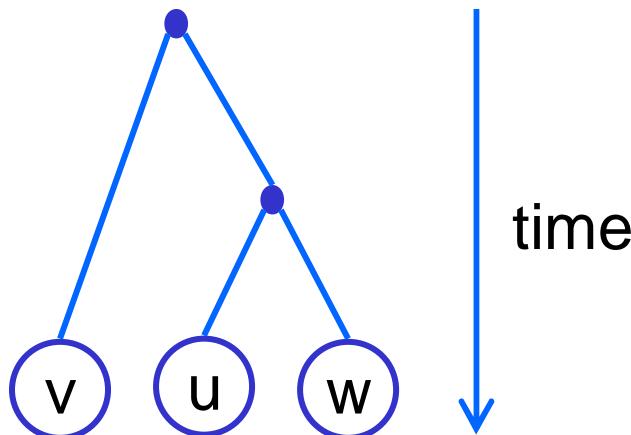
ultrametric distances

$$d(x,z) \leq \max \{ d(x,y), d(y,z) \}$$

$\Leftrightarrow$  three point condition

$$d(u,v) = d(v,w) \geq d(u,w)$$

for some ordering of  $u,v,w$



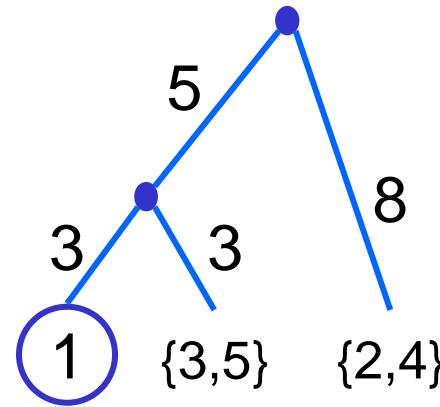
(uniform) molecular clock

# distance based

# ultrametric

sort according to distance to node 1

	1	2	3	4	5
1	-	16	6	16	6
2		-	16	8	16
3			-	16	2
4				-	16
5					-

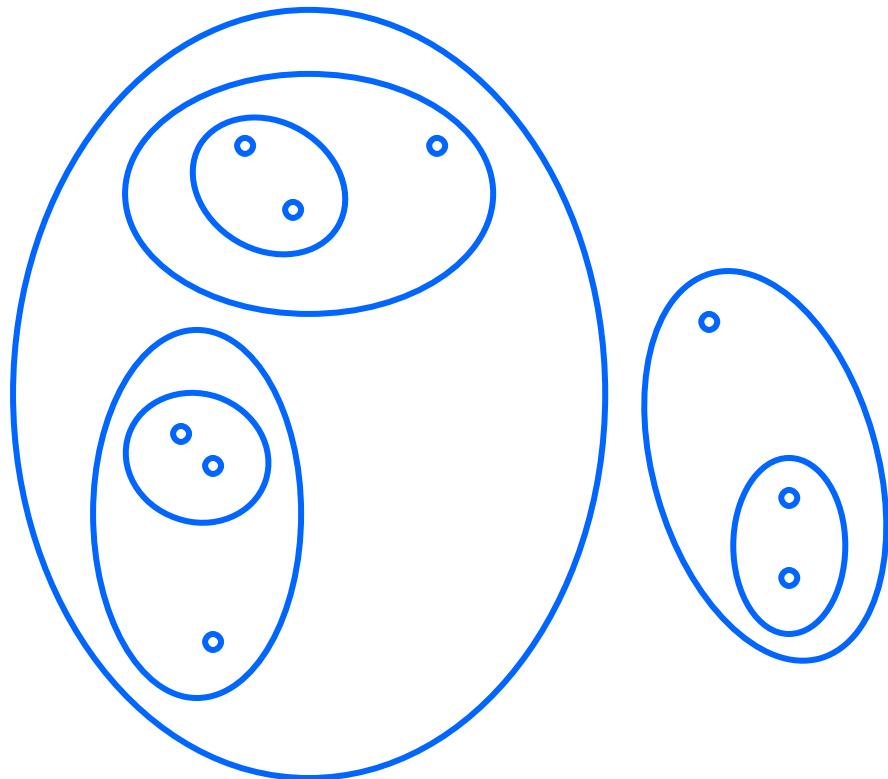


recursively solve for  $\{3,5\}$  and  $\{2,4\}$  (easy)

- ✓ conceptually simple, but
- ✓ tricky to implement efficiently
- ✓ can be adapted to the additive case
- ✓ even to character methods [says Gusfield]  
combinatorial rather than numeric
- ✓ UPGMA more general: clustering

distance based

clustering



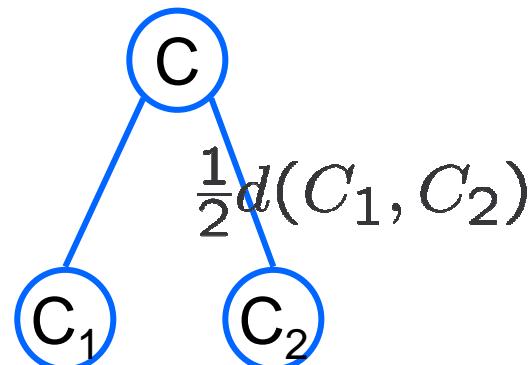
# distance based

# UPGMA

unweighted pair group method using arithmetic averages  
based on clustering

$$d(C_1, C_2) = \frac{1}{|C_1| \cdot |C_2|} \sum_{\substack{x \in C_1 \\ y \in C_2}} d(x, y)$$

- choose minimal distance clusters
- join them
- compute new distances



## distance based

## UPGMA

new distances from old

$$C = C_1 \cup C_2$$

$$d(C, C') = \frac{1}{|C| \cdot |C'|} \sum_{\substack{x \in C \\ y \in C'}} d(x, y)$$

$$d(C, C') = \frac{|C_1|}{|C|} d(C_1, C) + \frac{|C_2|}{|C|} d(C_2, C)$$

# UPGMA

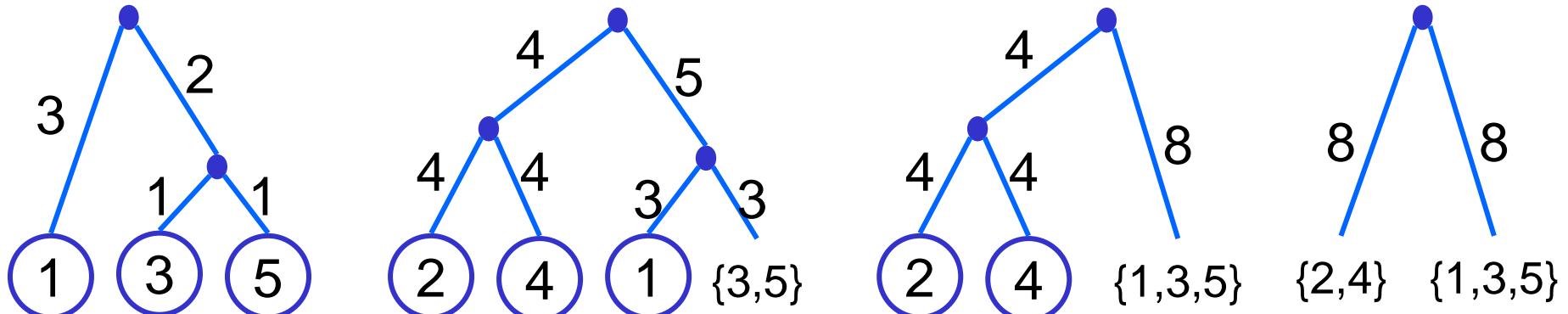
example

	1	2	3	4	5
1	-	16	6	16	6
2		-	16	8	16
3			-	16	2
4				-	16
5					-

	1	2	4	35
1	-	16	16	6
2		-	8	16
4			-	16
35				-

	2	4	135
2	-	8	16
4		-	16
135			-

	24	135
24	-	16
135		-



## distance based

## heuristics

comparing distances: matrix  $\mathbf{M}$  vs. tree  $\mathbf{d}$

$$\rho(M, d) = \sum_{xy} (M(x, y) - d(x, y))^2$$

optimization problem

- $\mathbf{M}$  given (matrix)
- $\mathbf{d}$  from suitable tree (to be found)

search ‘tree space’  
genetic algorithms, etc.