#### Protein bioinformatics: evolution

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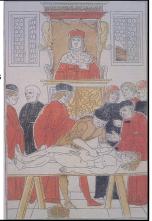
#### **Outline**

Sean Prigge described properties of amino acids, and an example of a multiple sequence alignment (globins).

Today we will discuss amino acid properties, and protein relatedness from an evolutionary perspective.

#### Outline

- 1. Pairwise alignment of proteins
- 2. Scoring matrices: how related are amino acids?
- 3. Multiple sequence alignment of proteins
- 4. From multiple sequence alignment to phylogenetic tree



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#### Pairwise sequence alignment is the most fundamental operation of bioinformatics

- It is used to decide if two proteins are related structurally or functionally
- It is used to identify domains or motifs that are shared between proteins
- It is the basis of BLAST searching

#### Pairwise alignments in the 1950s

β-corticotropin (sheep) Corticotropin A (pig)

ala gly glu asp asp glu asp gly ala glu asp glu

Oxytocin Vasopressin

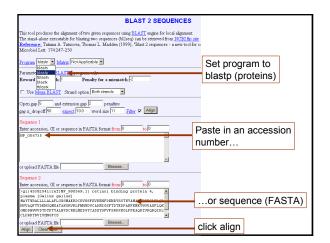
CYIQNCPLG CYFQNCPRG

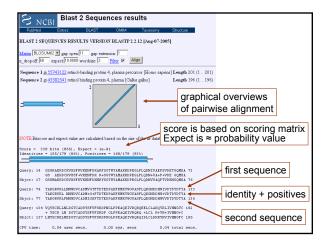
#### Pairwise alignment: BLAST 2 sequences

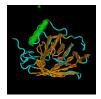
- Go to http://www.ncbi.nlm.nih.gov/BLAST
- Choose BLAST 2 sequences (bl2seq)
- In the program,
  - [1] choose blastp for proteins
  - [2] paste in your accession numbers (or use FASTA format)
  - [3] select optional parameters
    - --3 BLOSUM and 3 PAM matrices
    - --gap creation and extension penalties --filtering

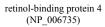
    - --word size
  - [4] click "align"











 $\beta\text{-lactoglobulin}$ (P02754)

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Pairwise alignment
The process of lining up two or more sequences to achieve maximal levels of identity (and conservation, in the case of amino acid sequences) for the purpose of assessing the degree of similarity and the possibility of homology.

#### **Definitions**

#### Homology

Similarity attributed to descent from a common ancestor.

#### Identity

The extent to which two (nucleotide or amino acid) sequences are invariant.

RBP: 26 RVKENFDKARFSGTWYAMAKKDPEGLFLQDNIVA 59
+ K++ ++ + GTW++MA + L + A
glycodelin: 23 QTKQDLELPKLAGTWHSMAMA-TNNISLMATLKA 55

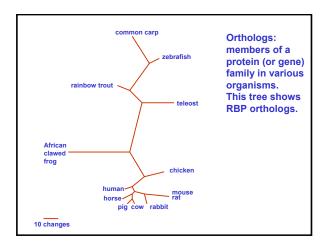
#### **Definitions: two types of homologs**

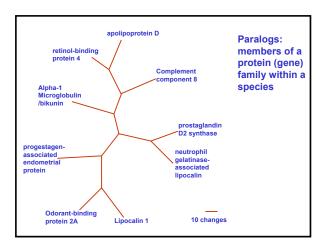
#### Orthologs

Homologous sequences in different species that arose from a common ancestral gene during speciation; may or may not be responsible for a similar function.

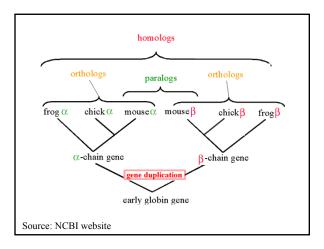
#### **Paralogs**

Homologous sequences within a single species that arose by gene duplication.





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#### **Definitions**

#### **Similarity**

The extent to which nucleotide or protein sequences are related. It is based upon identity plus conservation.

#### Identity

The extent to which two sequences are invariant.

#### Conservation

Changes at a specific position of an amino acid sequence that preserve the physico-chemical properties of the original residue.

Pairwise alignment of retinol-binding protein 4 and β-lactoglobulin: explaining the dots and dashes 1 MKWVWALLLLAAWAAAERDCRVSSFRVKENFDKARFSGTWYAMAKKDPEG 50 RBP . ||| | : |. . . | : .||||.:| : MKCLLLALALTCGAQALIVT..QTMKGLDIQKVAGTWYSLAMAASD. 44 lactoglobulin Very Somewhat similar similar (two dots) (one dot) **Identity** (bar) . | | | | 136 QCLVRTPEVDDEALEKFD .. 178 lactoglobulin Internal Terminal gap gap

#### **Gaps**

- Positions at which a letter is paired with a null are called gaps.
- · Gap scores are typically negative.
- Since a single mutational event may cause the insertion or deletion of more than one residue, the presence of a gap is ascribed more significance than the length of the gap.
- In BLAST, it is rarely necessary to change gap values from the default.

# Pairwise alignment of retinol-binding protein from human (top) and rainbow trout (*O. mykiss*): two closely related proteins

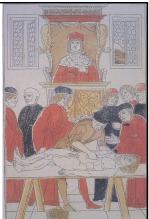
#### General approach to pairwise alignment

- Choose two sequences
- Select an algorithm that generates a score
- Allow gaps (insertions, deletions)
- Score reflects degree of similarity
- Alignments can be global (Needleman and Wunsch, 1970) or local (Smith and Waterman, 1981)
- Estimate probability that the alignment occurred by chance


# Calculation of an alignment score Range of Alignment ATTGTCAAAGACTTGAGCTGATGCAT GGCAGACATGA-CTGACAAGGGTATCG Mismatch S= Σ(identities, mismatches) - Σ (gap penalties) Score = Max(S) Source: http://www.ncbi.nlm.nih.gov/Education/BLASTinfo/Alignment\_Scores2.html

#### Outline

- 1. Pairwise alignment of proteins
- 2. Scoring matrices: how related are amino acids?
- 3. Multiple sequence alignment of proteins
- 4. From multiple sequence alignment to phylogenetic tree



### How do we decide what scores to assign in pairwise alignments?

- Zuckerkandl and Pauling (1965) made a multiple sequence alignment of hemoglobin and myoglobin from primates, horse, cattle, pig, lamprey, and carp. They made a "scoring matrix."
- Margaret Dayhoff and colleagues (1960s, 1970s) studied dozens of families of proteins to create scoring matrices that describe the relationship of well-conserved (or poorly-conserved) protein families.

### Multiple sequence alignment of glyceraldehyde 3-phosphate dehydrogenases

fly GARKVIISAP SAD.APM.F VCGVMLDAVK POMKVVSNAS CTTMCLAPLA plant GARKVIISAP SAD.APM.F VMGVMHEKYD NSLKIISAAS CTTMCLAPLA plant bacterium GARKVVIISAP SAD.APM.F VMGVMHEKYD NSLKIISAAS CTTMCLAPLA plant bacterium GARKVVITAPS SAD.APM.F VMGVMEKTT SPHMOIVSRAS CTTMCLAPLA GARKVVITAPS S.TAPM.F VMGVMEEKTT SLKIVSSAS CTTMCLAPLA archaeon GADKVLISAP PRODEVVGL VYGVMHEBYD GE.DVVSNAS CTTMCLAPLA prodevision of the company of the c

Studying conserved (and nonconserved) residues in closely related families may reveal "rules" for amino acid substitutions accepted by natural selection

### Multiple sequence alignment of human lipocalin paralogs

lipocalin 1 odorant-binding protein 2a progestagen-assoc. endo. apolipoprotein D retinol-binding protein neutrophil gelatinase-ass. prostaglandin D2 synthase alpha-1-microglobulin complement component 8

Studying conserved (and nonconserved) residues in distantly related families is also informative

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# PAM matrices: Point-accepted mutations

PAM matrices are based on global alignments of closely related proteins.

The PAM1 is the matrix calculated from comparisons of sequences with no more than 1% divergence.

Other PAM matrices are extrapolated from PAM1.

All the PAM data come from closely related proteins (>85% amino acid identity)

#### Dayhoff's 34 protein superfamilies

<u>Protein</u>	PAMs per 100 million years per 100 aa residues
lg kappa chain	37
kappa casein	33
luteinizing hormone b	30
lactalbumin	27
complement component 3	27
epidermal growth factor	26
proopiomelanocortin	21
pancreatic ribonuclease	21
haptoglobin alpha	20
serum albumin	19
phospholipase A2, group	IB 19
prolactin	17
carbonic anhydrase C	16
hemoglobin ά	12
hemoglobin β	12

#### Dayhoff's 34 protein superfamilies

<u>Protein</u>	PAMs per 100 million years per 100 aa residues
apolipoprotein A-II	10
lysozyme	9.8
gastrin	9.8
myoglobin	8.9
nerve growth factor	8.5
myelin basic protein	7.4
thyroid stimulating hormo	
parathyroid hormone	7.3
parvalbumin	7.0
trypsin	5.9
insulin	4.4
calcitonin	4.3
arginine vasopressin	3.6
adenylate kinase 1	3.2

#### Dayhoff's 34 protein superfamilies

#### **Protein**

### PAMs per 100 million years per 100 aa residues

triosephosphate isomerase 1	2.8
vasoactive intestinal peptide	2.6
glyceraldehyde phosph. dehydrogease	2.2
cytochrome c	2.2
collagen	1.7
troponin C, skeletal muscle	1.5
alpha crystallin B chain	1.5
glucagon	1.2
glutamate dehydrogenase	0.9
histone H2B, member Q	0.9
ubiquitin	0

### Dayhoff's numbers of "accepted point mutations": what amino acid substitutions occur in proteins?

	A	R	N	D	C	Q	E	G
	Ala	Arg	Asn	Asp	Cys	Gln	Glu	Gly
Α								
R	30							
N	109	17						
D	154	0	532					
С	33	10	0	0				
Q	93	120	50	76	0			
Е	266	0	94	831	0	422		
G	579	10	156	162	10	30	112	
Η	21	103	226	43	10	243	23	10

## Multiple sequence alignment of glyceraldehyde 3-phosphate dehydrogenases

	1				Į.
fly	GAKKVIISAP	SAD.APMF	VCGVNLDAYK	PDMKVVSNAS	CTTNCLAPLA
human	GAKRVIISAP	SAD.APMF	VMGVNHEKYD	NSLKIISNAS	CTTNCLAPLA
plant	GAKKVIISAP	SAD.APMF	VVGVNEHTYQ	PNMDIVSNAS	CTTNCLAPLA
bacterium	GAKKVVMTGP	SKDNTPMF	VKGANFDKY.	AGQDIVSNAS	CTTNCLAPLA
yeast	GAKKVVITAP	SS.TAPMF	VMGVNEEKYT	SDLKIVSNAS	CTTNCLAPLA
archaeon	GADKVLISAP	PKGDEPVKQL	VYGVNHDEYD	GE.DVVSNAS	CTTNSITPVA
fly	KVINDNFEIV	EGLMTTVHAT	TATQKTVDGP	SGKLWRDGRG	AAQNIIPAST
human	KVIHDNFGIV	EGLMTTVHAI	TATQKTVDGP	SGKLWRDGRG	ALQNIIPAST
plant	KVVHEEFGIL	EGLMTTVHAT	TATQKTVDGP	SMKDWRGGRG	ASQNIIPSST
bacterium	KVINDNFGII	EGLMTTVHAT	TATQKTVDGP	SHKDWRGGRG	ASQNIIPSST
yeast	KVINDAFGIE	EGLMTTVHSL	TATQKTVDGP	SHKDWRGGRT	ASGNIIPSST
archaeon	KVLDEEFGIN	AGQLTTVHAY	TGSQNLMDGP	NGKP.RRRRA	AAENIIPTST
fly	GAAKAVGKVI	PALNGKLTGM	AFRVPTPNVS	VVDLTVRLGK	GASYDEIKAK
human	GAAKAVGKVI	PELNGKLTGM	AFRVPTANVS	VVDLTCRLEK	PAKYDDIKKV
plant	GAAKAVGKVL	PELNGKLTGM	AFRVPTSNVS	VVDLTCRLEK	GASYEDVKAA
bacterium	GAAKAVGKVL	PELNGKLTGM	AFRVPTPNVS	VVDLTVRLEK	AATYEQIKAA
yeast	GAAKAVGKVL	PELQGKLTGM	AFRVPTVDVS	VVDLTVKLNK	ETTYDEIKKV
archaeon	GAAQAATEVL	PELEGKLDGM	AIRVPVPNGS	ITEFVVDLDD	DVTESDVNAA
					•

#### The relative mutability of amino acids Asn 134 His 66 65 Ser 120 Arg Asp 106 Lys 56 Glu 102 Pro 56 Ala 100 Gly 49 Thr 97 Tyr 41 lle 96 Phe 41 Met 94 Leu 40 Gln 93 Cys 20 Val 74 Trp 18

#### Normalized frequencies of amino acids Arg\* 4.1% Gly 8.9% Ala 8.7% Asn 4.0% Leu\* 8.5% Phe 4.0% Gln 3.8% Lys 8.1% Ser\* 7.0% 3.7% lle Val 6.5% His 3.4% Thr 5.8% Cys 3.3% Pro 5.1% Tyr 3.0% Glu 5.0% Met† 1.5% Asp 4.7% Trp† 1.0% blue\*=6 codons; red†=1 codon

		Seco	nd letter	1	E
	U	С	Α	G	
U	UUU Phe UUC Leu UUA Leu	UCU UCC UCA UCG	UAU Tyr UAC Stop UAG Stop	UGU Cys UGC Stop UGG Trp	DCAG
С	CUU CUC CUA CUG	CCU CCC CCA CCG	CAU His CAA GIn	CGU CGC CGA CGG	∪ C ≪ G
A	AUU AUC AUA AUG Met	ACU ACC ACA ACG	AAU Asn AAC AAA AAA Lys	AGU AGC Ser AGA AGG Arg	GUCAG
G	GUU GUC GUA GUG	GCU GCC GCA GCG	GAU Asp GAC Asp GAA GIU	GGU GGC GGA GGG	UCAG

### Dayhoff's numbers of "accepted point mutations": what amino acid substitutions occur in proteins?

	A	R	N	D	C	Q	Е	G
	Ala	Arg	Asn	Asp	Cys	Gln	Glu	Gly
A								
R	30							
N	109	17						
D	154	0	532					
С	33	10	0	0				
Q	93	120	50	76	0			
Е	266	0	94	831	0	422		
G	579	10	156	162	10	30	112	
Н	21	103	226	43	10	243	23	10

#### Dayhoff's PAM1 mutation probability matrix

	A	R	N	D	C	Q	Е	G	Н
	Ala	Arg	Asn	Asp	Cys	Gln	Glu	Gly	His
Α	9867	2	9	10	3	8	17	21	2
R	1	9913	1	0	1	10	0	0	10
N	4	1	9822	36	0	4	6	6	21
D	6	0	42	9859	0	6	53	6	4
С	1	1	0	0	9973	0	0	0	1
Q	3	9	4	5	0	9876	27	1	23
Е	10	0	7	56	0	35	9865	4	2
G	21	1	12	11	1	3	7	9935	1
Н	1	8	18	3	1	20	1	0	9912
I	2	2	3	1	2	1	2	0	0

Each element of the matrix shows the probability that an original amino acid (top) will be replaced by another amino acid (side)

#### **Substitution Matrix**

A substitution matrix contains values proportional to the probability that amino acid *i* mutates into amino acid *j* for all pairs of amino acids.

Substitution matrices are constructed by assembling a large and diverse sample of verified pairwise alignments (or multiple sequence alignments) of amino acids.

Substitution matrices should reflect the true probabilities of mutations occurring through a period of evolution.

The two major types of substitution matrices are PAM and BLOSUM.

## PAM matrices: Point-accepted mutations

PAM matrices are based on global alignments of closely related proteins.

The PAM1 is the matrix calculated from comparisons of sequences with no more than 1% divergence.

Other PAM matrices are extrapolated from PAM1.

All the PAM data come from closely related proteins (>85% amino acid identity)

### Dayhoff's PAM0 mutation probability matrix: the rules for extremely slowly evolving proteins

PAM0	A	R	N	D	С	Q	E
	Ala	Arg	Asn	Asp	Cys	Gln	Glu
A	100%	0%	0%	0%	0%	0%	0%
R	0%	100%	0%	0%	0%	0%	0%
N	0%	0%	100%	0%	0%	0%	0%
D	0%	0%	0%	100%	0%	0%	0%
С	0%	0%	0%	0%	100%	0%	0%
Q	0%	0%	0%	0%	0%	100%	0%
E	0%	0%	0%	0%	0%	0%	100%
G	0%	0%	0%	0%	0%	0%	0%

Top: original amino acid Side: replacement amino acid

### Dayhoff's PAM2000 mutation probability matrix: the rules for very distantly related proteins

PAM∞	A	R	N	D	С	Q	Е	G
	Ala	Arg	Asn	Asp	Cys	Gln	Glu	Gly
A	8.7%	8.78	8.78	8.78	8.78	8.78	8.78	8.7
R	4.1%	4.18	4.18	4.18	4.18	4.18	4.18	4.1
N	4.0%	4.0%	4.0%	4.0%	4.0%	4.0%	4.0%	4.0
D	4.7%	4.78	4.78	4.78	4.78	4.78	4.78	4.7
С	3.3%	3.38	3.38	3.38	3.38	3.38	3.38	3.3
Q	3.8%	3.88	3.88	3.8%	3.8%	3.8%	3.8%	3.8
Е	5.0%	5.0%	5.0%	5.0%	5.0%	5.0%	5.0%	5.0
G	8.9%	8.98	8.98	8.98	8.98	8.98	8.98	8.9

Top: original amino acid Side: replacement amino acid

# 

Side: replacement amino acid

A	2	ı																		
R	-2	6	ı																	
N.	-2	0	2	ı																
D	0	-1	2	4																
C	-2	-4	-4	-5	12	ı					_									
0	-2	1	1	2	-5	4	ı				Ρ/	417	125	v	ιοί	g c	ac	วร		
E E	0	-1	1	3	-5	2	4	ı			61	٠.	in	n r	na	tri	v			
G	1	-3	0	1	-3	-1	0	5	l		30	,01	••••	y '	III		^			
H	-1	2	2	1	-3	3	1	-2	6	1										
11 T	-1	-2	-2	-2	-2	-2	-2	-3	-2	5										
L	-2	-3	-3	-4	-6	-2	-3	-4	-2	-2	6	l								
K	-1	3	1	-4	-5	1	-0	-2	-2	-2	-3	5								
M	-1	0	-2	-3	-5	-1	-2	-3	-2	2	4	0	6							
F	-3	-4	-3	-6	-4	-5	-5	-5	-2	1	2	-5	0	9						
P	1	0	0	-1	-3	0	-1	0	0	-2	-3	-1	-2	-5	6	l				
S	1	0	1	0	0	-1	0	1	-1	-1	-3	0	-2	-3	1	2	ĺ			
T	1	-1	0	0	-2	-1	0	0	-1	0	-2	0	-1	-3	0	1	3/		\	
W	-6	2	-4	-7	-8	-5	-7	-7	-3	-5	-2	-3	-4	0	-6	-2	-5	17	1)	
Y	-3	-4	-2	-4	0	-4	-4	-5	0	-1	-1	-4	-2	7	-5	-3	-3	0	10	
v	0	-2	-2	-2	-2	-2	-2	-1	-2	4	2	-2	2	-1	-1	-1	0	-6	-2	4
	A	R	N	D	C	0	E	G	Н	I	L	K	M	F	P	S	Т	W	Y	V

# Why do we go from a mutation probability matrix to a log odds matrix?

 We want a scoring matrix so that when we do a pairwise alignment (or a BLAST search) we know what score to assign to two aligned amino acid residues.

		38 bits (866), Expect = 1e-91 = 155/179 (86%), Positives = 168/179 (93%)
Query:	14	GSGRAERDCRVSSFRVKENFDKARFSGTWYAMAKKDPEGLFLQDNIVAEFSVDETGQMSA 73 GS AERDCRVSSF-VKENFDK R-SGTWYAMAKKDPEGLFLODN+VA+F-VDE GOMSA
Sbjct:	17	GSSMAERDCRVSSFWKENFOR MAGGINTAMARKDPEGLFLQDNVVAQFTVDENGQMSA 76
Query:	74	
Sbjot:	77	TAKGRURL NNUDUCADH-G-FTDTEDPAKFKNEYUGVASFLQKGNDDHW-VDTDYDTYA TAKGRURLFNNUDUCADHIGSFTDTEDPAKFKNEYUGVASFLQKGNDDHWVVDTDYDTYA 136
Query:	134	VQYSCRLLNLDGTCADSYSFVFSRDPNGLPPEAQKIVRQRQEELCLARQYRLIVHNGYC 192
	***	+ YSCR LW DGTCADSYSFVFSRDP GLPPEAQKIVRQRQ +LCL R+YR+IVHNG+C

 Logarithms are easier to use for a scoring system. They allow us to sum the scores of aligned residues (rather than having to multiply them).

# How do we go from a mutation probability matrix to a log odds matrix?

• The cells in a log odds matrix consist of an "odds ratio":

the probability that an alignment is authentic the probability that the alignment was random

The score S for an alignment of residues a,b is given by:

 $S(a,b) = 10 log_{10} (M_{ab}/p_b)$ 

As an example, for tryptophan,

 $S(a,tryptophan) = 10 log_{10} (0.55/0.010) = 17.4$ 

#### Normalized frequencies of amino acids

		R	N	D	С	0	Е	G	Н	Y	T	K	М	F	Р	s	Т	W	Y	v	Arg Asn	4.1% 4.0%
Α	A 13	K 6	N 9	D	5	Q 8	E Q	12	H 6	8	L 6	7.	7	4	11	11	11	2 2	4	V 9	Phe	4.0%
R	3	17	4	3	2	5	3	2	6	3	2	9	4	1	4	4	3	7	2	2	Gln	3.8%
N	4	4	6	7	2	5	6	4	6	3	2	5	3	2	4	5	4	2	3	3	lle	3.7%
D	5	4	8	11	1	7	10	5	6	3	2	5	3	1	4	5	5	1	2	3	His	3.4%
С	2	1	1	1	52	1	1	2	2	2	1	1	1	1	2	3	2	1	4	2		
Q	3	5	5	6	1	10	7	3	7	2	3	5	3	1	4	3	3	1	2	3	Cys	3.3%
E	5	4	7	11	1	9	12	5	6	3	2	5	3	1	4	5	5	1	2	3	Tyr	3.0%
G	12	5	10	10	4	7	9	27	5	5	4	6	5	3	8	11	9	2	3	7	Met	1.5%
Н	2	5	5	4	2	7	4	2	15	2	2	3	2	2	3	3	2	2	3	2	Trp	1.0%
I	3	2	2	2	2	2	2	2	2	10	6	2	6	5	2	3	4	1	3	9	пр	1.0 /0
L	6	4	10	3	2	6	4	3	5	15	34	4	20	13	5	4	6	6	7	13		
K	6	18	10	8	0	10	8	5	8	5	3	24	6	2	6	8	8	4	3	5		
F	2	1	2	1	1	1	1	1	3	5	6	1	4	32	1	2	2	4	20	3		
P	7	5	5	4	3	5	4	5	5	3	3	4	3	2	20	6	5	1	20	4		
S	9	6	8	7	7	6	7	9	6	5	4	7	5	3	9	10	9	4	4	6		
T	8	5	6	6	4	5	5	6	4	6	4	6	5	3	6	8	1	2	1	6		
W	0	2	0	0	0	0	0	0	1	0	1	0	0	1	0	1	1	55	1	0		
Y	1	1	2	1	3	1	1	1	3	2	2	1	2	15	1	2	2	3	ı	2		
V	7	4	4	4	4	4	4	5	4	15	10	4	10	5	5	5	7	2	4	17		

# What do the numbers mean in a log odds matrix?

 $S(a,tryptophan) = 10 log_{10} (0.55/0.010) = 17.4$ 

A score of +17 for tryptophan means that this alignment is 50 times more likely than a chance alignment of two Trp residues.

 $\begin{array}{l} S(a,b) = 10 \ log_{10} \ (M_{ab}/p_b) \\ S(a,b) = 17 \\ Probability of replacement \ (M_{ab}/p_b) = x \\ Then \\ 17 = 10 \ log_{10} \ x \\ 1.7 = log_{10} \ x \\ 10^{1.7} = x \\ 50 = x \end{array}$ 

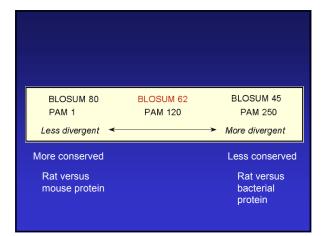

# What do the numbers mean in a log odds matrix?

A score of –10 indicates that the correspondence of two amino acids in an alignment that accurately represents homology (evolutionary descent) is one tenth as frequent as the chance alignment of these amino acids.

A score of 0 is neutral.

A score of +2 indicates that the amino acid replacement occurs 1.6 times as frequently as expected by chance.

Α	7																			
R	-10	9		,																
N	-7	-9	9		,															
D	-6	-17	-1	8																
C	-10	-11	-17	-21	10							_								
Q	-7	-4	-7	-6	-20	9						Р	Αľ	И1	0 l	og	00	ds	3	
E	-5	-15	-5	0	-20	-1	8		,						gı					
G	-4	-13	-6	-6	-13	-10	-7	7				5	CO	1111	ıgı	IIId	ıuı	X		
Н	-11	-4	-2	-7	-10	-2	-9	-13	10											
I	-8	-8	-8	-11	-9	-11	-8	-17	-13	9										
L	-9	-12	-10	-19	-21	-8	-13	-14	-9	-4	7									
K	-10	-2	-4	-8	-20	-6	-7	-10	-10	-9	-11	7								
M	-8	-7	-15	-17	-20	-7	-10	-12	-17	-3	-2	-4	12							
F	-12	-12	-12	-21	-19	-19	-20	-12	-9	-5	-5	-20	-7	9						
P	-4	-7	-9	-12	-11	-6	-9	-10	-7	-12	-10	-10	-11	-13	8					
S	-3	-6	-2	-7	-6	-8	-7	-4	-9	-10	-12	-7	-8	-9	-4	7				
T	-3	-10	-5	-8	-11	-9	-9	-10	-11	-5	-10	-6	-7	-12	-7	-2	8			
W	-20	-5	-11	-21	-22	-19	-23	-21	-10	-20	-9	-18	-19	-7	-20	-8	-19	13		
Y	-11	-14	-7	-17	-7	-18	-11	-20	-6	-9	-10	-12	-17	-1	-20	-10	-9	-8	10	
v	-5	-11	-12	-11	-9	-10	-10	-9	-9	-1	-5	-13	-4	-12	-9	-10	-6	-22	-10	
	A	R	N	D	С	0	E	G	Н	I	L	K	M	F	P	S	T	W	Y	V



# Comparing two proteins with a PAM1 matrix gives completely different results than PAM250! Consider two distantly related proteins. A PAM40 matrix is not forgiving of mismatches, and penalizes them

severely. Using this matrix you can find almost no match.

hsrbp, 136 CRLINIDGTC
btlact, 3 CALLALALTC

\* \* \* \* \* \*\*

#### A PAM250 matrix is very tolerant of mismatches.

24.7% identity in 81 residues overlap; Score: 77.0; Gap frequency: 3.7% rbp4 26 RVKENFDKARFSGTWAMAKKDPEGLFLQDMIVAEFSVDETGQMSATAKGRVELLNNNDV btlact 21 QTMKGLDIQKVAGTWYSLAMASD: JSLDAQSAPLRVVYBELKFTFEGDLETLLQXWEN

rbp4 86 --CADMVGTFTDTEDPAKFKM btlact 80 GECAQKKIIAEKTKIPAVFKI \*\* \* \*\* \*\*

#### PAM: "Accepted point mutation"

- Two proteins with 50% identity may have 80 changes per 100 residues. Why? Because any residue can be subject to back mutations.
- Proteins with 20% to 25% identity are in the "twilight zone" and may be statistically significantly related.
- PAM or "accepted point mutation" refers to the "hits" or matches between two sequences (Dayhoff & Eck, 1968)

#### Outline

- 1. Pairwise alignment of proteins
- 2. Scoring matrices: how related are amino acids?
- 3. Multiple sequence alignment of proteins
- 4. From multiple sequence alignment to phylogenetic tree




# Multiple sequence alignment: definition • a collection of three or more protein (or nucleic acid) sequences that are partially or completely aligned · homologous residues are aligned in columns across the length of the sequences residues are homologous in an evolutionary sense · residues are homologous in a structural sense Multiple sequence alignment: properties • not necessarily one "correct" alignment of a protein family • protein sequences evolve... • ...the corresponding three-dimensional structures of proteins also evolve • may be impossible to identify amino acid residues that align properly (structurally) throughout a multiple sequence alignment • for two proteins sharing 30% amino acid identity. about 50% of the individual amino acids are superposable in the two structures Multiple sequence alignment: features • some aligned residues, such as cysteines that form disulfide bridges, may be highly conserved • there may be conserved motifs such as a transmembrane domain

there may be conserved secondary structure features
 there may be regions with consistent patterns of

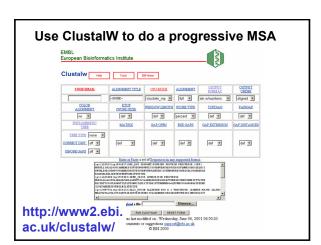
insertions or deletions (indels)

# Multiple sequence alignment: uses • MSA is more sensitive than pairwise alignment to detect homologs · BLAST output can take the form of a MSA, and can reveal conserved residues or motifs • Population data can be analyzed in a MSA (PopSet) • A single query can be searched against a database of MSAs (e.g. PFAM) Regulatory regions of genes may have consensus sequences identifiable by MSA Multiple sequence alignment: methods There are two main ways to make a multiple sequence alignment: (1) Progressive alignment (Feng & Doolittle). We will illustrate this using ClustalW. (2) Iterative approaches Multiple sequence alignment: methods Example of MSA using ClustalW: two data sets Five distantly related lipocalins (human to *E. coli*) Five closely related RBPs When you do this, obtain the sequences of interest in the FASTA format! (You can save them in a Word document)

# The input for ClustalW: a group of sequences (DNA or protein) in the FASTA format >LYSC\_TRAVT/19-146 KIFERGELARTLKKLGLDOYRGYSLANGVCLAKUESGYNTEATNYNPGDESTDYGIFQIN SRYWCNNOKTPGAVDACHISCSALLQNNIADAVACARRVVSDPQGIRAWVAWRNHCQNKD VSOVYKCC >LYSC1\_CAPRLY-1-127 KVFERCELARTLKKLGLDDYKGYSLANGLCLTKUESGYNTKAINYNPGSESTDYGIFQIN SRFUCHDOKTPDAYDGGFWSCSELNENDIEKAVACAKHIVSE-QGITAWVAWKSHCRDHD VSSYVEGC >LYSC2\_CARDR/1-128 KVWERCALARKLKELGHDGYRGVSLANGMCLTKUESDYNTDAINYNPSSESTDYGIFQIN SRYWCHNOKTPHAVNNGGTNCNVLLEDDITKAVQCAKRVVRDPQGVRAWVAWKHCEGHD VSQYVEGC >LYSC2\_CONCHY/16-142 KVYDRCELARALKASGMDGYAGNSLPHWVCLSKUESSYNTQAITNRNT-DGSTDYGIFQIN SRYWCHOGRTFGANNVGIRCSQLLTADLTVAIRCAKRVVLDPNGIGAWVAWRHCHCQNGD LBSYVAGC >LYSC1\_PIG/1-126 KVYDRCEFARILKRSGMDGYBGVSLANGWCLAKWESDFNTKAINEN--VGSTDYGIFQIN SRYWCHOGKTFRANNACHISCKYLLDDDLSQDIECAKRVVRDPQGIKAWVAWRHCHCQNKD VSQYTIRCC >LYSC1\_RTA/19-146 KYYERCGFARTLKKNGHSGYYGVSLADWVCLAGHESNYNTQARNYNPGDGSTDYGIFQIN

SRYWCNDGKTPRAKNACGIPCSALLQDDITQAIQCAKRVVRDPQGIRAWVAWQRHCKNRD

LSGYIRNC



# Feng-Doolittle MSA occurs in 3 stages [1] Do a set of global pairwise alignments (Needleman and Wunsch's dynamic programming algorithm) [2] Create a guide tree [3] Progressively align the sequences

# Progressive MSA stage 1 of 3: generate global pairwise alignments CLESTAL F [1.82] Baltiple Sequence Alignments CLESTAL F [1.82] Baltiple Sequence Alignments Sequence (1.01) (1.010) (1.01

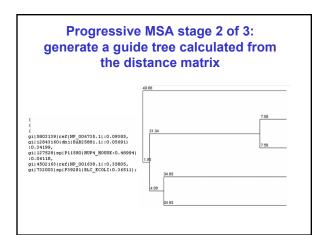
#### **Progressive MSA stage 1 of 3:** generate global pairwise alignments Sequence 2: gi 6174963 | sp | Q00724 | RETB MOUS Sequence 3: gi 132407 | sp | P04916 | RETB RAT Sequence 4: gi 89271 | pir | h39486 Sequence 5: gi | 132403 | sp | P18902 | RETB BOVIN Start of Pairwise alignments Aligning... Sequences (1:2) Aligned. Score: 84 five closely Sequences (1:3) Aligned. Score: 84 related lipocalins Sequences (1:4) Aligned. Score: 91 Sequences (1:5) Aligned. Score: 92 best Sequences (2:3) Aligned. Score: 99 score Sequences (2:4) Aligned. Score: 86 Sequences (2:5) Aligned. Score: 85 Sequences (3:4) Aligned. Score: 85 Sequences (3:5) Aligned. Score: 84

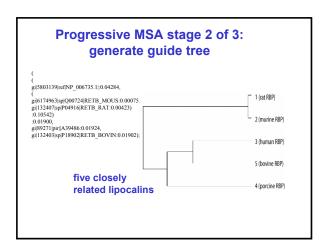
Sequences (4:5) Aligned. Score: 96

# Number of pairwise alignments needed For n sequences, (n-1)(n) / 2For 5 sequences, (4)(5) / 2 = 10

#### Feng-Doolittle stage 2: guide tree

- · Convert similarity scores to distance scores
- A tree shows the distance between objects
- Use UPGMA (defined below)
- ClustalW provides a syntax to describe the tree





#### Feng-Doolittle stage 3: progressive alignment

- Make a MSA based on the order in the guide tree
- · Start with the two most closely related sequences
- Then add the next closest sequence
- Continue until all sequences are added to the MSA
- Rule: "once a gap, always a gap."

# Progressive MSA stage 3 of 3: progressively align the sequences following the branch order of the tree

CLUSTAL W (1.81) multiple sequence alignment

gi 5803139 ref NP_006735.1	HKWVWALLLLAAWAAAERDCRVSSFRVKENFDKARFSG	38
gi 12843160 dbj BAB25881.1	MEWVWALVLLAALGOGSAERDCRVSSFRVKENFDKARFSG	40
gi 4502163 ref NP_001638.1	MVMLLLLLSALAGLFGAREGQAFHLGECPNPPVQENFIVNKYLG	44
gi 732003 sp P39281 BLC ECOLI	MRLLPLVAAATAAFLVVACSSPTPPRGVTVVNNFDAKRYLG	41
g1 127528 sp P11590 MUP4_MOUSE	MKLLLCLGLTLVCIHAEEATSKGONLNVEKING	33
	**	
g1 5803139 ref NP_006735.1	TWYAMAKKDPEGLFLQDNIVAEFSVDETGQMSATAKGRVRLLNNWDVCAD	88
gi 12843160 dbj BAB25881.1	LWYAIAKKDPEGLFLQDNIIAEFSVDEKGHMSATAKGRVRLLSNWEVCAD	90
gi 4502163 ref NP 001638.1	RWYEIEK-IPTTFENGRCIQANYSLMENGKIKVLNQELRADGTVNQIE	91
g1 732003 sp P39281 BLC_ECOLI	TWYEIARFDHRFERGLEKVTATYSLRDDGGLNVINKGYNP-DRGMWQQSE	90
gi 127528 sp P11590 MUP4_MOUSE	EWFSILLASDKREK-IEEMGSMRVFVEHIHVLENSLAFKFHTVIDGECSE	82
	***	
gi 5803139 ref NP_006735.1	MVGTFTDTEDPAKFKMKYWGVASFLQKGNDDHWIVDTDYDTYAVQYSCRL	138
g1 12843160 db3 EAE25881.1	HVGTFTDTEDPAKFKHKYWGVASFLQRGNDDHWIIDTDYDTFALQYSCRL	140
gi 4502163 ref NP 001638.1	GEATPVNLTEPAKLEVKFSWFMPSAPYWILATDYENYALVYSC	134
g1 732003 sp P39281 BLC_ECOLI	GKAYFTGAPTRAALKVSFFGPFYGGYNVIALDREYR-	126
gi 127528 sp P11590 MUP4 MOUSE	IFLVADETERAGEYSVHYDGFNTFTILETDYDNYIHFHLIN-	123
gi 5803139 ref NP_006735.1	LNLDGTCADSYSFVFSRDPNGLPPEAQKIVRQRQEELCLARQYRLIVHNG	188
gi 12843160 dbj BAB25881.1	QNLDGTCADSYSFVFSRDPNGLSPETRRLVRQRQEELCLERQYRWIEHNG	190
gi 4502163 ref NP 001638.1	TCIIQLFHVDFAWILARNPN-LPPETVDSLKNILTSNNIDVKKHTVTDQV	183
g1 732003 sp P39281 BLC_ECOLI	HALVCGPDRDYLWILSRTPT-ISDEVKQEMLAVATREGFDVSKFIWVQQP	175
gi 127528 sp P11590 MUP4 MOUSE	EKBGKTFQLMELYGRKADLNSDIKEKFVKLCEENGIIKENIIDLTKTN	171
	and the second of the second	
gi 5803139 ref NP 006735.1	YCDGRSERNLL 199	
gi 12843160 db3 BAB25881.1	YCQSRPSRNSL 201	
g1 4502163 ref NP_001638.1	NCPKLS 189	
gi 732003 sp P39281 BLC ECOLI	08 177	

#### Clustal W alignment of 5 closely related lipocalins

CLUSTAL W (1.82) multiple sequence alignment

gi | 89271 | pir | | A39486 gi | 132403 | sp | P18902 | RETB BOVIN gi | 5803139 | ref | NP\_006735.1 | gi | 6174963 | sp | Q00724 | RETB MOUS gi | 132407 | sp | P04916 | RETB\_RAT

gi | 89271 | pir | | A39486 gi | 132403 | sp | P18902 | RETB\_BOVIN gi | 5803139 | ref | NP\_06735. 1 | gi | 6174963 | sp | Q00724 | RETB\_MOUS gi | 132407 | sp | P04916 | RETB\_RAT

gi | 89271 | pir | | A39486 gi | 132403 | sp | P18902 | RETB\_BOVIN gi | 5803139 | ref | NP\_006735. 1 | gi | 6174963 | sp | Q00724 | RETB\_MOUS gi | 132407 | sp | P04916 | RETB\_RAT MEWWALIVILAALGSAQAERDCRVSSFRVKENFDKARFSGTKYAMAKKOP 50
EBCGVSSFRVKENFDKARFSGTKYAMAKKOP 32
MINWALILILAAN--AAABERCCRVSSFRVKENFDKARFSGTKYAMAKKOP 32
MINWALILILAAN-SAABERCCRVSSFRVKENFDKARFSGTKYAMAKKOP 50
MEWWALIVILAALGGGSAERDCRVSSFRVKENFDKARFSGTKYATAKKOP 50
MEWWALIVILAALGGGSAERDCRVSSFRVKENFDKARFSGTKYATAKKOP 50

RGLELGDBIYABEFSUERUMHSATAKKEVELLINMEWICADMVOTFTUTED 100
EGIFLCDBIYABEFSUERUMHSATAKKEVELLINMEWICADMVOTFTUTED 100
EGIFLCDBIYABEFSUERUMHSATAKKEVELLINMEWICADMVOTFTUTED 90
EGIFLCDBIYABEFSUERUMHSATAKKEVELLISMEVICADMVOTFTUTED 100
EGIFLCDBIYABEFSUERUMHSATAKKEVELLISMEVICADMVOTFTUTED 100
EGIFLCDBIYABEFSUERUMHSATAKKEVELLISMEVICADMVOTFTUTED 100

PAKFRHKYHOVASFLQKENDDHHIIDTDYDTYAAQYSCRLQNLDDTCADS 150
PAKFRHKYHOVASFLQKENDDHHIIDTDYDTFAVOYSCRLANLDDTCADS 152
PAKFRHKYHOVASFLQKENDDHHIIDTDYDTAVOYSCRLANLDDTCADS 152
PAKFRHKYHOVASFLQKENDDHIIDTDYDTPALOSCRLANLDDTCADS 150
PAKFRHKYHOVASFLQKENDDHHIIDTDYDTPALOSCRLANLDDTCADS 150

\* asterisks indicate identity in a column

### Additional features of ClustalW improve its ability to generate accurate MSAs

- Individual weights are assigned to sequences; very closely related sequences are given less weight, while distantly related sequences are given more weight
- Scoring matrices are varied dependent on the presence of conserved or divergent sequences, e.g.:

PAM20 80-100% id PAM60 60-80% id PAM120 40-60% id PAM350 0-40% id

• Residue-specific gap penalties are applied

#### Outline

- 1. Pairwise alignment of proteins
- 2. Scoring matrices: how related are amino acids?
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#### Four stages of phylogenetic analysis

Molecular phylogenetic analysis may be described in four stages:

- [1] Selection of sequences for analysis
- [2] Multiple sequence alignment
- [3] Tree building
- [4] Tree evaluation


#### Stage 1: Use of DNA, RNA, or protein

For some phylogenetic studies, it may be preferable to use protein instead of DNA sequences. With DNA, one can also study synonymous versus nonsynonymous mutations, noncoding DNA, pseudogenes, etc.

#### Stage 2: Multiple sequence alignment

The fundamental basis of a phylogenetic tree is a multiple sequence alignment.

(If there is a misalignment, or if a nonhomologous sequence is included in the alignment, it will still be possible to generate a tree.)

Consider the following alignment of 13 orthologous retinol-binding proteins.

				1	2	3
		1		· • • • • • • • • • • • • • • • • • • •		▼ 50
	ccrbp	MLRLCIALCV	LATCWAQDFL	ESNTTVKQDČ		LVSNITVKQD
fish	drrbp	MLRLCIAVCV	IA		TCWAQDC	
11011	omrbp				~~~~~SDC	
	sarbp		~~~~~MT	RMLRYVVALC	LLAVSWAQDC	
	mmrbp		MEWVW	.ALVLLAA	LGGGSAERDC	
	rnrbp	~~~~~~~	~~~~MEWVW	.ALVLLAA	LGGGSAERDC	
	btrbp	~~~~~~~				
other	ssrbp		~~~~MEWVW		LGSAQAERDC	
omer		~~~~~~~	~~~~MKWVW			
	hsrbp	~~~~~~~	~~~~MEWVW		WAAAERDC LGSGRGERDC	
				RALLLIALAF		
	ggrbp			LGL. LIALGE		
	xrrub	1.1	~~~~DELECT.V	LALL LIMING	TOSCIMENTO	RVDNEEVMKD
		51 4	5			6 L L 7
	ccrbp		VAVAKKDEVG	LFLLDNVVAN	FEVORDGTMT	ATATGRUITI.
	drrbp	FNRTRYOGTW		LFLLDNIVAN		
fish	omrbp	FDRSRYTGRW		LFLLDNYVAO		
	sarbp	FDKTRYAGTW		LFLIDNIVAO		
	mmrbp	FDKARFSGLW	YATAKKDPEG	LFLODNIIAE	<b>FSVDRKGHMS</b>	ATAKGRURLL
	rnrbp	FDKARFSGLW			FSVDEKGHMS	
	btrbp	FDKARFAGTW	YAMAKKDPEG	LFLODNIVAE	FSVDENGHMS	ATAKGRVRLL
other	ssrbp	FDKARFSGTW	YAMAKKDPEG	LFLQDNIVAE	FSVDENGHMS	ATAKGRVRLL
Oniei	ecrbp	FDKARFSGTW	YAMAKKDPEG	LFLQDNIVAE	FSVDEYGQMS	ATAKGRVRLL
	hsrbp	FDKARFSGTW	YAMAKKDPEG	LFLQDNIVAE	FSVDETGQMS	ATAKGRVRLL
	ocrbp	FDKARFAGTW	YAMAKKDPEG	LFLQDNIVAE	FSVDENGHMS	ATAKGRVRLL
	ggrbp	FDKNRYSGTW		LFLQDNVVAQ		
	xlrbp	FNKERYAGVW	YAVAKKDPEG	LFLLDNIAAN	FKIEDNGKTT	ATAKGRVRIL
			9	↓1	0 11	
		101 8				150
	ccrbp	NNMENCANME		RFKMKYWGAA		
fish	drrbp			KFKMKYWGAA		
	omrbp	NNWEMCANME		KFKMRYWGAA		
	sarbp			KFRMRYWGAA		
	mmrbp	SNWEVCADMV		KFKMKYWGVA		
	rnrbp			KFKMKYWGVA		
	btrbp			KFKMKYWGVA		
other	ssrbp			KFKMKYWGVA		
	ecrbp	NNWDVCADMV		KFKMKYWGVA KFKMKYWGVA		
				KFKMKYWGVA		
	ocrbp			KEKMKYWGVA		
	ggrop			KYRMKYHGAL		
	· vrrnh	DECEMBER OF STREET	GILLWINDPA	ALLOW I DOME	PLEASING LIDER	WAADTDIIII

		1		1 ↓	2	<b>J3</b> 50	
	ccrbp	MLRLCIALCV					
fish	drrbp	MLRLCIAVCV					
11311	omrbp	~~~~~~~					
l	sarbp	~~~~~~~					
	mmrbp	~~~~~~~					
	rnrbp btrbp						
l	ssrbp	~~~~~~~					
other	ecrbp	~~~~~~~					
	hsrbp	~~~~~~	~~~~MKWVW	.ALLLLAA	WAAAERDC	RVSSFRVKEN	
	ocrbp	~~~~~~~	$\sim\!\sim\!\sim\!\sim\!\texttt{MEWVW}$	.ALVLLAA	LGSGRGERDC	RVSSFRVKEN	
	ggrbp	~~~~~~~					
l	xlrbp	~~~~~~~	~~~~MERKV	LGL.LIALGF	LGSCLAEKNC	RVDNFEVMKD	
I							
۔ ا			مامندان بمسم		-1:		
I 20	me pos	sitions of the	e multiple	sequence	aligriment	are	
invariant (arrow 2). Some positions distinguish fish RBP from all other RBPs (arrow 3).							

#### Stage 2: Multiple sequence alignment

- [1] Confirm that all sequences are homologous
- [2] Adjust gap creation and extension penalties as needed to optimize the alignment
- [3] Restrict phylogenetic analysis to regions of the multiple sequence alignment for which data are available for all taxa (delete columns having incomplete data—even if a gap occurs in only one taxon).
- [4] In this example, note that four RBPs are from fish, while the others are vertebrates that evolved more recently.

#### Stage 3: Tree-building methods

We will discuss two tree-building methods: distance-based and character-based.

Distance-based methods involve a distance metric, such as the number of amino acid changes between the sequences, or a distance score. Examples of distance-based algorithms are UPGMA and neighbor-joining.

#### Stage 3: Tree-building methods

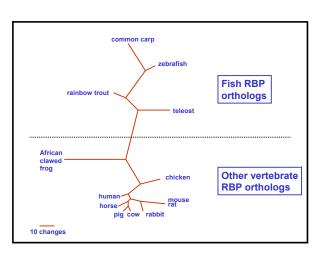
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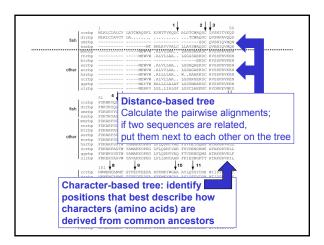
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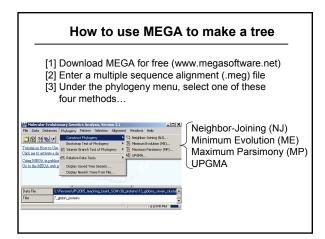
Character-based methods include maximum parsimony and maximum likelihood. Parsimony analysis involves the search for the tree with the fewest amino acid (or nucleotide) changes that account for the observed differences between taxa.

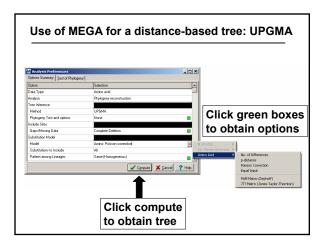
#### Stage 3: Tree-building methods

We can introduce distance-based and character-based tree-building methods by referring to a tree of 13 orthologous retinol-binding proteins, and the multiple sequence alignment from which the tree was generated.









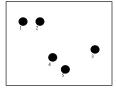
# Use of MEGA for a distance-based tree: UPGMA | Comparison | Conclusion | Boold | SON(38) prediction | 13 (piddons, creen, district) | Prince | Pri

# Use of MEGA for a distance-based tree: UPGMA | March 3 Tree Deplorer (S.) Presence 3 (2008) good loop, load | SCH(3) genders(13, globans, severa, shad all typidams areas, shad all typidams areas,

# UPGMA is unweighted pair group method using arithmetic mean

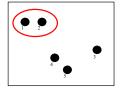
#### Tree-building methods: UPGMA

Step 1: compute the pairwise distances of all the proteins. Get ready to put the numbers 1-5 at the bottom of your new tree.



#### Tree-building methods: UPGMA

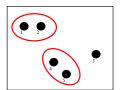
Step 2: Find the two proteins with the smallest pairwise distance. Cluster them.

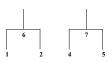


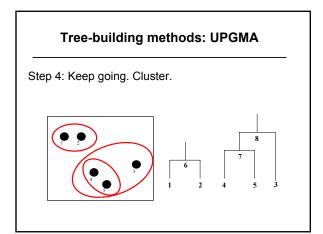


#### Tree-building methods: UPGMA

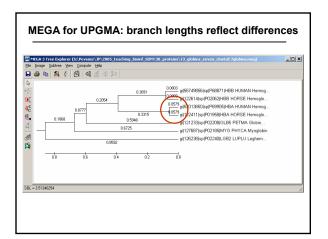
Step 3: Do it again. Find the next two proteins with the smallest pairwise distance. Cluster them.

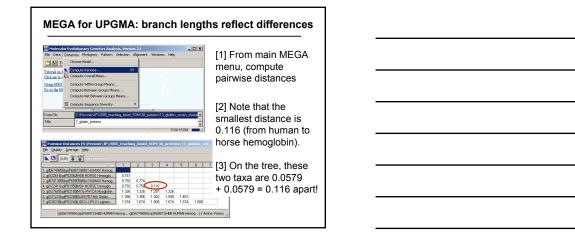






# Tree-building methods: UPGMA Step 4: Last cluster! This is your tree.





#### **Stage 4: Evaluating trees**

The main criteria by which the accuracy of a phylogentic tree is assessed are consistency, efficiency, and robustness. Evaluation of accuracy can refer to an approach (e.g. UPGMA) or to a particular tree.

#### **Stage 4: Evaluating trees: bootstrapping**

Bootstrapping is a commonly used approach to measuring the robustness of a tree topology. Given a branching order, how consistently does an algorithm find that branching order in a randomly permuted version of the original data set?

#### Stage 4: Evaluating trees: bootstrapping

Bootstrapping is a commonly used approach to measuring the robustness of a tree topology. Given a branching order, how consistently does an algorithm find that branching order in a randomly permuted version of the original data set?

To bootstrap, make an artificial dataset obtained by randomly sampling columns from your multiple sequence alignment. Make the dataset the same size as the original. Do 100 (to 1,000) bootstrap replicates. Observe the percent of cases in which the assignment of clades in the original tree is supported by the bootstrap replicates. >70% is considered significant.

