# Chapter III: Pairwise Sequence Alignment

#### **Presentations use info from:**

Jonathan Pevsner, Ph.D.
http://bioinfbook.org
pevsner@kennedykrieger.org
Bioinformatics and Functional Genomics
(3<sup>rd</sup> edition, ©2015 John Wiley & Sons, Ltd.)
You may use this PowerPoint for teaching purposes

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# What will you learn?

To define homology as well as orthologs and paralogs

To explain how PAM (accepted point mutation) matrices are derived

To contrast the utility of PAM and BLOSUM scoring matrices.

To define dynamic programming and explain how global and local pairwise alignments are performed

To perform pairwise alignment of protein or DNA sequences at the NCBI website

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

- It is used to decide if two proteins (or genes) 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.
- It is used in the analysis of genomes.

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

# Protein alignment: often more informative than DNA alignment

Sequence alignment: protein sequences can be more informative than DNA

Protein is more informative (20 vs 4 characters);

Many amino acids share related biophysical properties

Codons are degenerate: changes in the third position

Often do not alter the amino acid that is specified

Protein sequences offer a longer "look-back" time

# Sequence alignment: protein sequences can be more informative than DNA

#### **Example:**

- ->searching for plant globins using human beta globin **DNA** yields no matches;
- ->searching for plant globins using human beta globin protein yields many matches.

#### Web BLAST



#### blastx

translated nucleotide ▶ protein

#### tblastn

protein ▶ translated nucleotide



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

#### Many times, DNA alignments are appropriate:

- >to study non-coding regions of DNA (e.g., introns or intergenic regions)
- database searching
- ➤ to study DNA polymorphisms
- relies on DNA analysis

#### **Definitions**

#### Homology

➤ Similarity attributed to descent from a common ancestor.

#### **Identity**

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

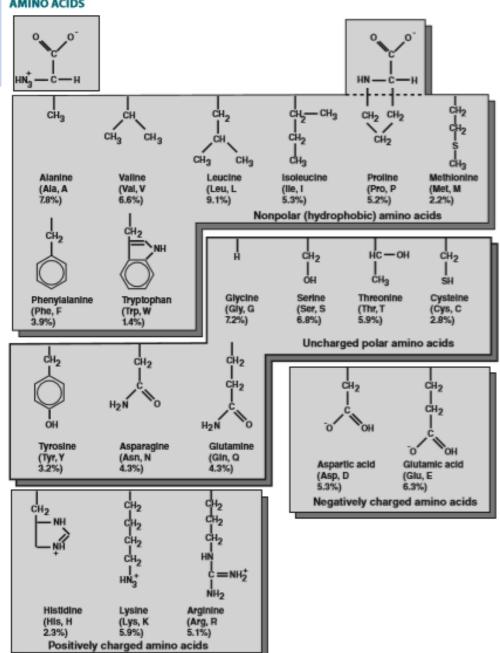
#### **Similarity**

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

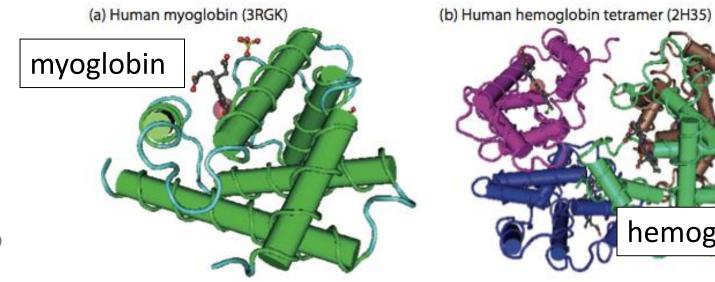
#### Conservation

Changes at a specific position of an amino acid or (less commonly, DNA) sequence that preserve the physio-chemical properties of the original residue.

#### BOX 3.2 STRUCTURES AND ONE- AND THREE-LETTER ABBREVIATIONS OF 20 COMMON AMINO ACIDS

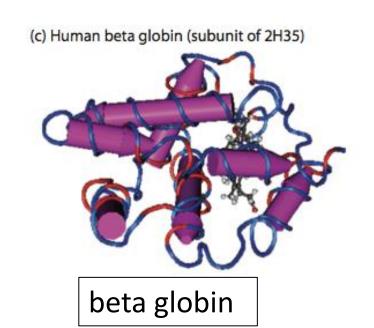


- basic amino acids (K, R, H)
- acidic amino acids (D, E)
- hydroxylated amino acids (S, T)
- hydrophobic amino acids(W, F, Y, L, I, V, M, A)



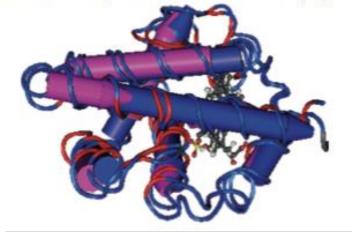
### Globin homologs

In general, three-dimensional structures diverge much more slowly than amino acid sequence identity between two proteins.

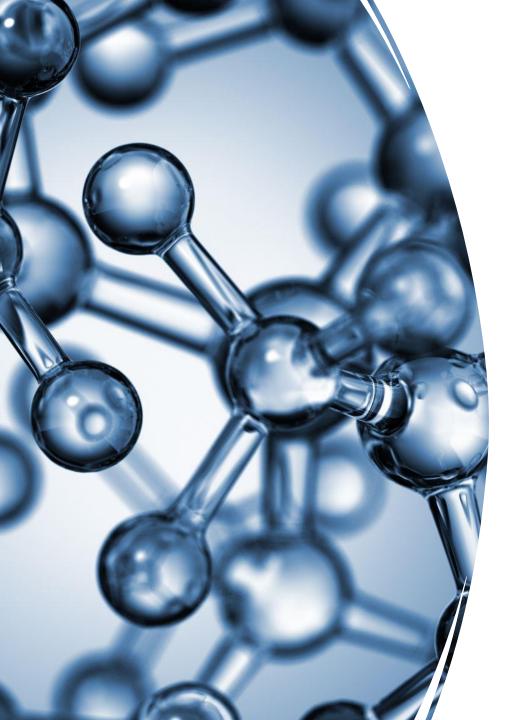


hemoglobin

(d) Pairwise alignment of beta globin and myoglobin



beta globin and myoglobin (aligned)



# Definitions: two types of homology

#### **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.

# Example of paralogs: HBA2 and HBA1 genes

#### HBA2 hemoglobin subunit alpha 2 [ Homo sapiens (human) ]

previous assembly

Gene ID: 3040, updated on 10-Oct-2020

105



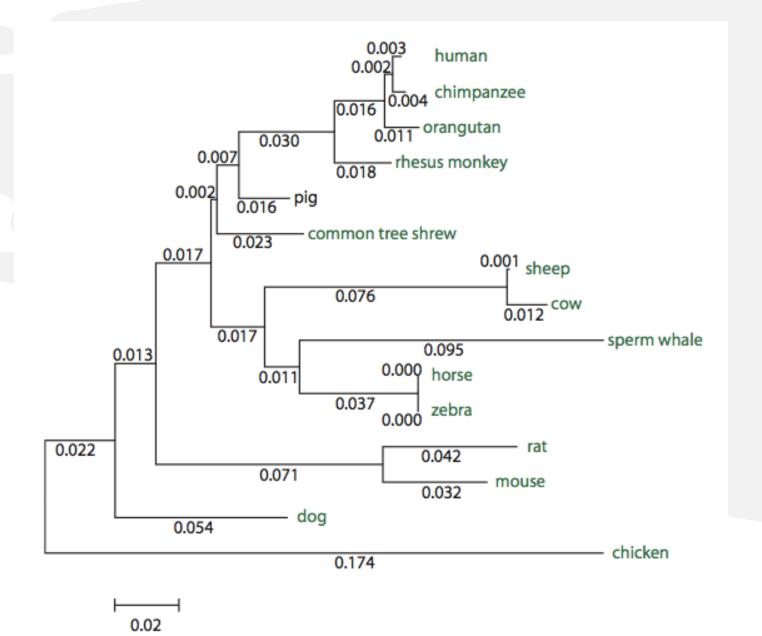


16

NC 000016.9 (222846..223709)

GRCh37.p13 (GCF 000001405.25)

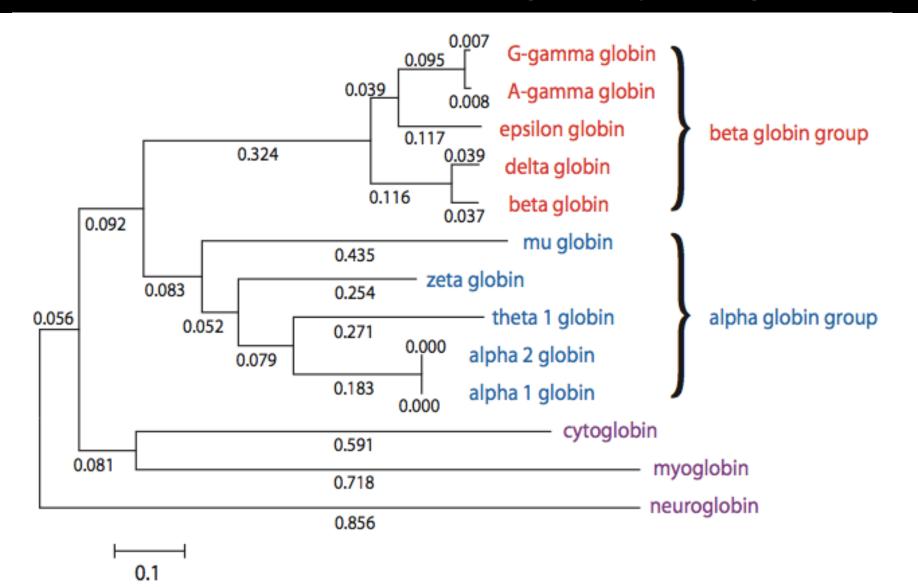
#### Myoglobin proteins: examples of orthologs





#### Paralogs: members of a gene (protein) family within a species

#### This tree shows human globin paralogs.

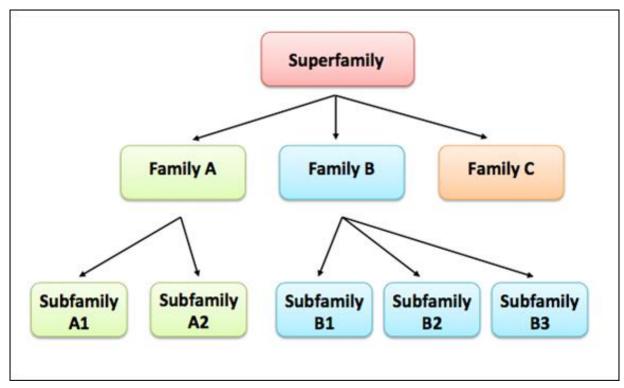


#### Why classify proteins?

- Proteins can be classified into groups according to sequence or structural similarity.
- These groups often contain well characterized proteins whose function is known.
- When a novel protein is identified, its functional properties can be proposed based on the group to which it is predicted to belong.

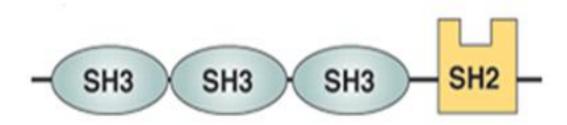
#### What are protein families?

A group of proteins that share a common evolutionary origin, reflected by their related functions and similarities in sequence or structure.



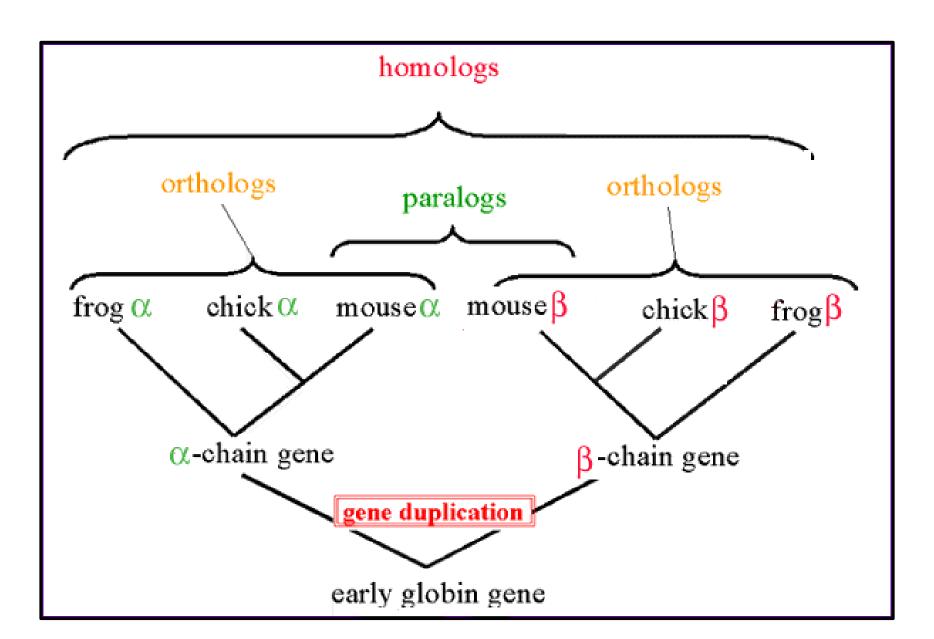
#### What are protein domains?

- Domains are distinct functional and/or structural units in a protein.
- They are responsible for a particular function or interaction, contributing to the overall role of a protein.
- Domains may exist in a variety of biological contexts, where similar domains can be found in proteins with different functions.



Domain composition of Nck

#### Orthologs and paralogs are often viewed in a single tree





# 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 or local
- ✓ Estimate probability that the alignment occurred by chance

Find BLAST from the home page of NCBI and select protein BLAST...

Basic Local Alignment Search

#### Popular Resources PubMed Bookshelf PubMed Central **BLAST** Nucleotide Genome SNP Gene Protein PubChem

#### Basic Local Alignment Search Tool

BLAST finds regions of similarity between biological sequences. The program compares nucleotide or protein sequences to sequence databases and calculates the statistical significance.

Learn more

The Basic Local Alignment Search Tool (BLAST) finds regions of local similarity between sequences. The program compares nucleotide or protein sequences to sequence databases and calculates the statistical significance of matches. BLAST can be used to infer functional and evolutionary relationships between sequences as well as help identify members of gene families.

#### ClusteredNR database on BLAST+

The ClusteredNR database is now available for BLAST+

Thu, 24 Aug 2023

More BLAST news...

#### **Web BLAST**

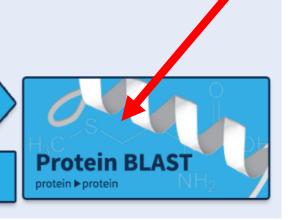


#### blastx

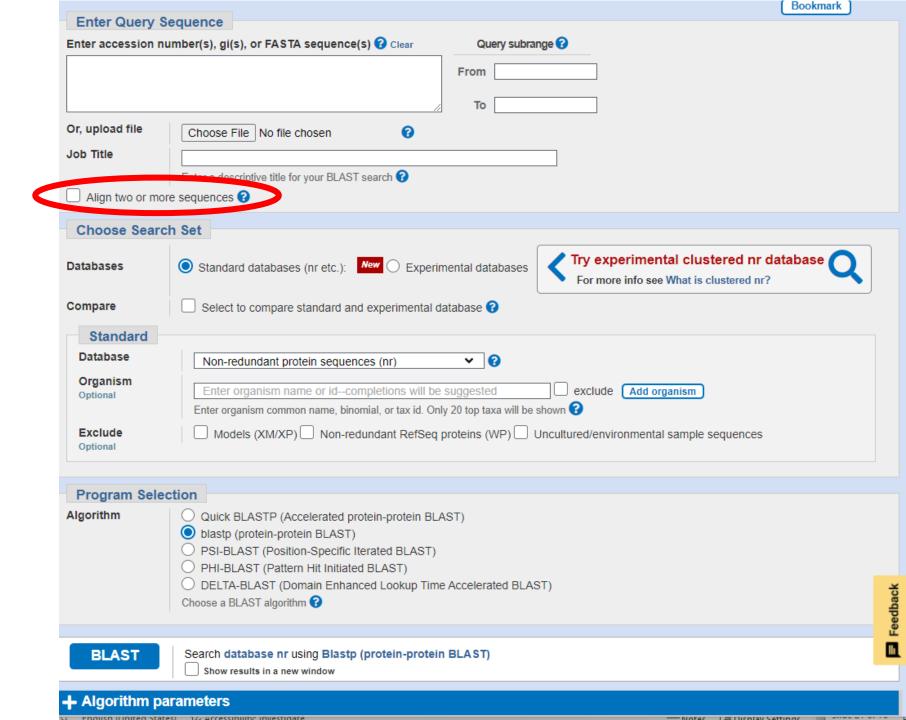
translated nucleotide ▶ protein

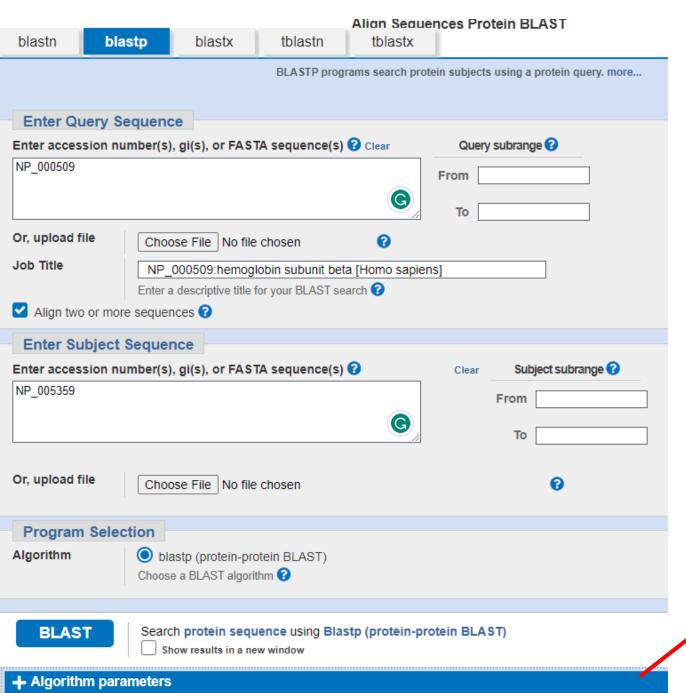
#### tblastn

protein ▶ translated nucleotide



 Choose align two or more sequences...





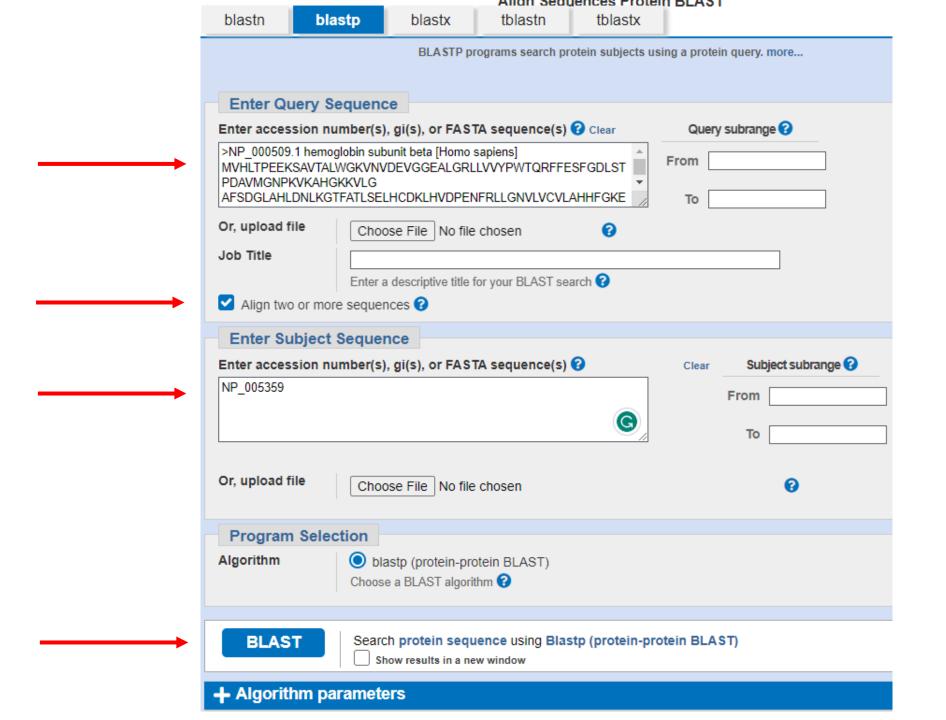
Enter the two sequences (as accession numbers or in the fasta format) and click **BLAST**.

NP\_000509 NP\_005359

Optionally select "Algorithm parameters" and note the matrix

option.

- Algorithm para	ameters
General Paran	anto-ra
General Paran	neters
Max target sequences	Select the maximum number of aligned sequences to display <b>?</b>
Short queries	Automatically adjust parameters for short input sequences ?
Expect threshold	0.05
Word size	3 <b>v</b> ?
Max matches in a query range	0
Scoring Paran	neters
Matrix	♦ BLOSUM45 ♥ ②
Gap Costs	Existence: 13 Extension: 3 🕶 🔞
Compositional adjustments	Conditional compositional score matrix adjustment ▼   ②
Filters and Ma	sking
Filter	Low complexity regions ?
Mask	☐ Mask for lookup table only �� ☐ Mask lower case letters ��
BLAST	Search protein sequence using Blastp (protein-protein BLAST)  Show results in a new window



# Pairwise alignment of human beta globin (the "query") and myoglobin (the "subject")

#### myoglobin isoform 1 [Homo sapiens]

```
Sequence ID: NP_001349775.1 Length: 154 Number of Matches: 1

See 12 more title(s) ➤ See all Identical Proteins(IPG)
```

DFGADAQGAMNKALELFRKDMASNY

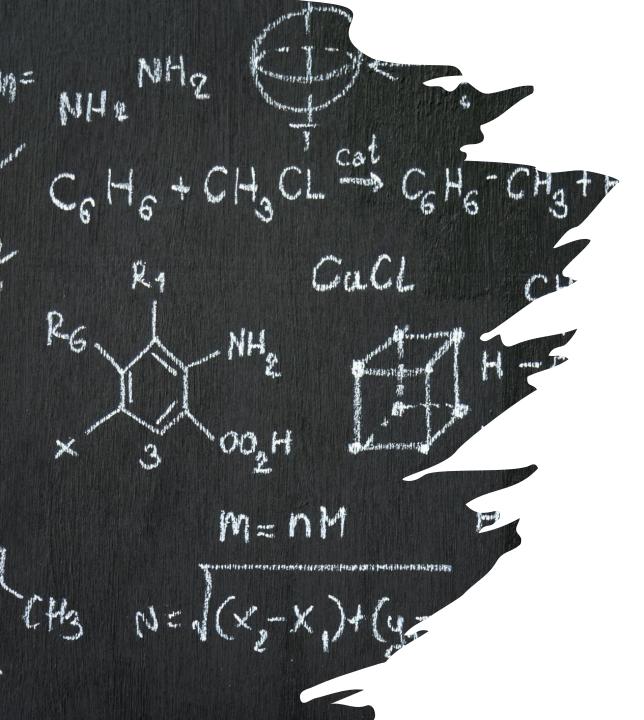
Sbjct

```
Range 1: 3 to 147 GenPept Graphics
                                                         ▼ Next Match ▲ Previous Match
 Score = 43.9 bits (102), Expect = 1e-09, Method: Composition-based stats.
 Identities = 37/145 (25%), Positives = 57/145 (39%), Gaps = 2/145 (1%)
Ouerv 4
          LTPEEKSAVTALWGKVNVD--EVGGEALGRLLVVYPWTQRFFESFGDLSTPDAVMGNPKV
                                  GELRL
                   V +WGKV D
Sbjct 3 LSDGEWQLVLNVWGKVEADIPGHGQEVLIRLFKGHPETLEKFDKFKHLKSEDEMKASEDL
Query 62
           KAHGKKVLGAFSDGLAHLDNLKGTFATLSELHCDKLHVDPENFRLLGNVLVCVLAHHFGK
        → K HG VL A
                                      L++HK++
Sbjct 63
           KKHGATVLTALGGILKKKGHHEAEIKPLAQSHATKHKIPVKYLEFISECIIQVLQSKHPG
          EFTPPVQAAYQKVVAGVANALAHKY 146
```

We'll examine the highlighted green region of the alignment in more detail.

#### How raw scores are calculated: an example

For a set of aligned residues, we assign scores based on matches, mismatches, gap open penalties, and gap extension penalties. These scores add up to the total raw score.



## Where do scores come from?

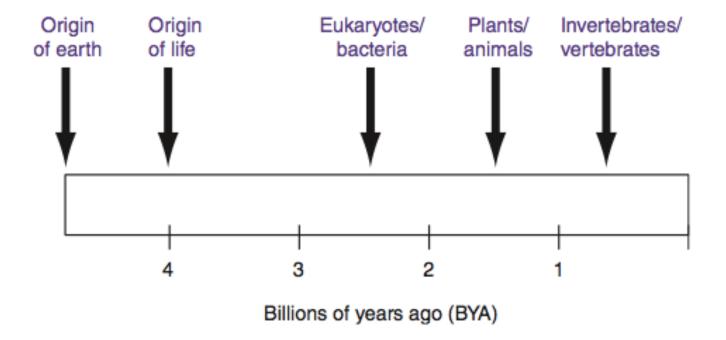
We'll examine scoring matrices. These are related to the properties of the 20 common amino acids.

### Gap

- 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. Thus, there are separate penalties for gap creation and gap extension.
- In BLAST, it is rarely necessary to change gap values from the default.

## Pairwise alignment and the evolution of life

- When two proteins (or DNA sequences) are homologous they share a common ancestor.
- We can infer the sequence of that ancestor.
- When we align globins from human and a plant we can imagine their common ancestor, a single celled organism that lived 1.5 billion years ago, and we can infer that ancient globin sequence.
- Through pairwise alignment we can look back in time at sequence evolution.

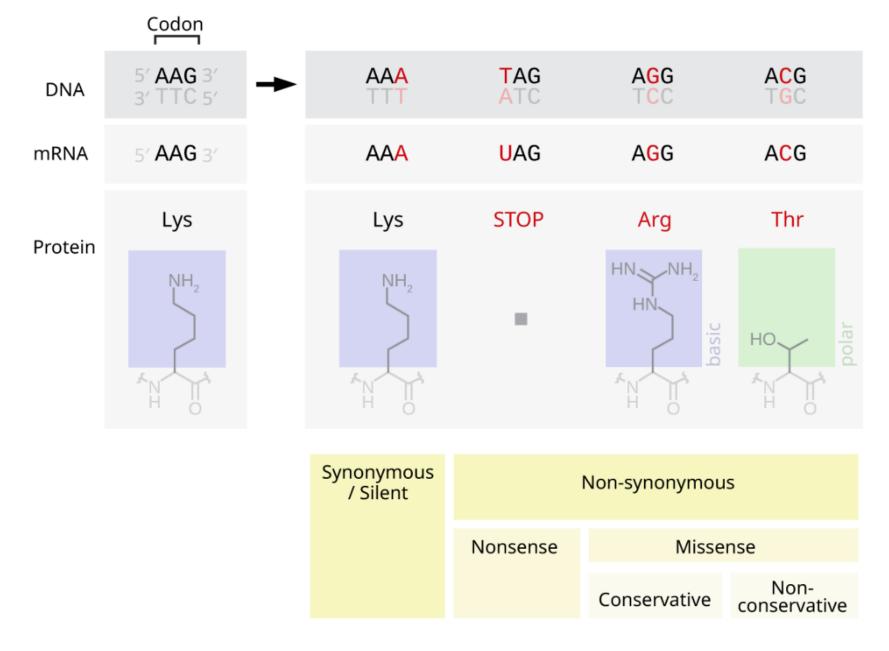


#### Scoring matrices -> Dayhoff model: 7 steps

Step 1: Accepted point mutations (PAMs) in protein families

PROTEIN	PAMS PER 100 MILLION YEARS
Immunoglobulin (Ig) kappa chain C region	37
Kappa casein	33
Epidermal growth factor	26
Serum albumin	19
Hemoglobin alpha chain	12
Myoglobin	8.9
Nerve growth factor	8.5
Trypsin	5.9
Insulin	4.4
Cytochrome c	2.2
Glutamate dehydrogenase	0.9
Histone H3	0.14
Histone H4	0.10

Margaret Dayhoff and colleagues developed scoring matrices in the 1960s and 1970s. They defined PAMs as "accepted point mutations." Some protein families evolve very slowly (e.g. histones change little over 100 million years); others (such as kappa casein) change very rapidly.



An example of point mutations at an amino acid site coding for lysine. The missense mutations may be classed as **point accepted mutations** if the **mutated protein is not rejected by natural selection** (Source: Wikipedia).

<u>Protein</u>	PAMs per	100 million	<u>years</u>
Ig kappa chain Kappa casein luteinizing hormone b lactalbumin complement compone epidermal growth fact proopiomelanocortin pancreatic ribonucleas haptoglobin alpha serum albumin phospholipase A2, gro prolactin carbonic anhydrase C Hemoglobin a Hemoglobin b	ent 3 cor	37 33 30 27 27 26 21 21 20 19 19 17 16 12	
0.0000		<del>- —</del>	

#### **Protein**

#### PAMs per 100 million years

Ig kappa chain Kappa casein	7
luteinizing hormone b lactalbumin complement component 3 epidermal growth factor proopiomelanocortin  3 2 3 3 3 2 3 3 3 3 3 3 3 3 3 3 3 3 3	7 7
pancreatic ribonuclease 2 haptoglobin alpha 20	
serum albumin I's phospholipase A2, group IB	score - 57.0 pits (130), Expect - 3e-07
prolactin carbonic anhydrase C Hemoglobin a Hemoglobin b	Query 61 NLYQRRPAI-AINNPYVPRTYYANPAVVRPHAQIPQRQYLPNSHPPTVVRLPNLHPSF 117 N Y RP++ A +PY+ ++R A I + Q +PN V +PSF

<u>Protein</u>	PAMs per I	00 million ye	<u>ars</u>
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 ho	ormone b	7.4	
parathyroid hormone		7.3	
parvalbumin		7.0	
trypsin		5.9	
insulin		4.4	
calcitonin		4.3	
arginine vasopressin		3.6	
adenylate kinase I		3.2	

<u>Protein</u>	PAMs per 100	<u>million</u>	<u>years</u>	
triosephosphate	isomerase I		2.8	
vasoactive intest	inal peptide		2.6	
glyceraldehyde p	hosph. dehydrogease	<u> </u>	2.2	
cytochrome c	1 / 0		2.2	
collagen			1.7	
troponin C, skel	etal muscle		1.5	
alpha crystallin E			1.5	
glucagon			1.2	
glutamate dehyd	rogenase		0.9	
histone H2B, me			0.9	
ubiquitin			0	

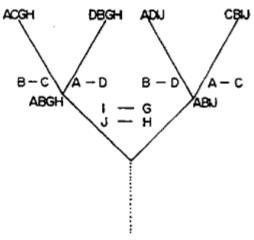
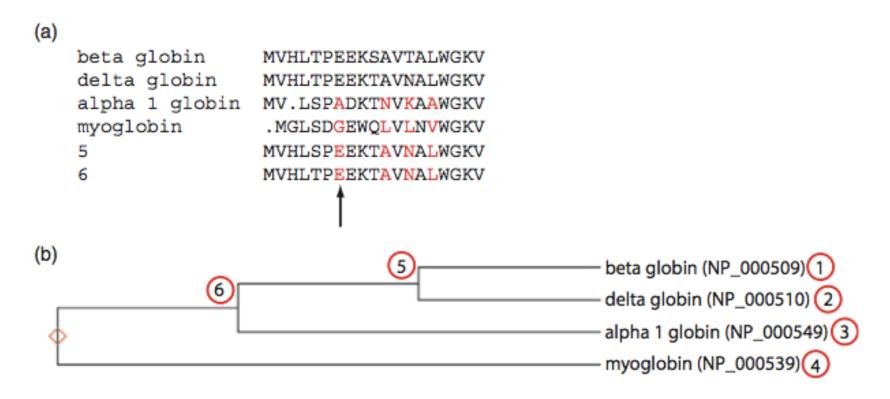


Figure 78. Simplified phylogenetic tree. Four "observed" proteins are shown at the top. Inferred ancestors are shown at the nodes. Amino acid exchanges are indicated along the branches.

	Α	В	С	D	6	н	I	J
А			1	1				
В			1	1				
С	1	1						
D	1	1						
G							1	
н					-			1
I					1			
J						1		

Figure 79. Matrix of accepted point mutations derived from the tree of Figure 78.

## Step 1: accepted point mutations are defined not by the pairwise alignment but with respect to the common ancestor



Dayhoff et al. evaluated amino acid changes. They applied an evolutionary model to compare changes such as 1 versus 2 not to each other but to an inferred common ancestor at position 5.

	Α	R	N	D	С	Q	Е	G	Н	I	L	K	M	F	P	S	T	W	Y	V
	Ala	Arg	Asn	Asp	Cys	Gln	Glu	Gly	His	Ile	Leu	Lys	Met	Phe	Pro	Ser	Thr	Trp	Tyr	Val
Α																				
R	30																			
N	109	17																		
D	154	0	532																	
С	33	10	0	0																
Q	93	120	50	76	0															
E	266	0	94	831	0	422														
G	579	10	156	162	10	30	112													
Н	21	103	226	43	10	243	23	10												
I	66	30	36	13	17	8	35	0	3											
L	95	17	37	0	У	75	15	17	40	253										
K	57	477	322	85	0	147	104	60	23	43	39									
M	29	17	0	0	0	20	7	7	0	57	207	90								
F	20	7	7	0	0	0	0	17	20	90	167	0	17							
P	345	67	27	10	10	93	40	49	50	7	43	43	4	7						
S	772	137	432	98	117	47	86	450	26	20	32	168	20	40	269					
T	590	20	169	57	10	37	31	50	14	129	52	200	28	10	73	696				
W	0	27	3	0	0	0	0	0	3	0	13	0	0	10	0	17	0			
Y	20	3	36	0	30	0	10	0	40	13	23	10	0	260	0	22	23	6		
V	365	20	13	17	33	27	37	97	30	661	303	17	77	10	50	43	186	0	17	
	Α	R	N	D	С	Q	Е	G	Н	I	L	K	M	F	P	S	T	W	Y	V
	Ala	Arg	Asn	Asp	Cys	Gln	Glu	Gly	His	Ile	Leu	Lys	Met	Phe	Pro	Ser	Thr	Trp	Tyr	Val

From a survey of 1572 observed substitutions, the original amino acid (columns) are compared to the changes (rows).

## Dayhoff model step 2 (of 7): Frequency of amino acids

TABLE 3.1 Normalized frequencies of amino acid. These values sum to 1. If the 20 amino acids were equally represented in proteins, these values would all be 0.05 (i.e., 5%); instead, amino acids vary in their frequency of occurrence.

Gly	0.089	Arg	0.041
Ala	0.087	Asn	0.040
Leu	0.085	Phe	0.040
Lys	0.081	Gln	0.038
Ser	0.070	lle	0.037
Val	0.065	His	0.034
Thr	0.058	Cys	0.033
Pro	0.051	Tyr	0.030
Glu	0.050	Met	0.015
Asp	0.047	Тгр	0.010

If 20 amino acids occurred in nature at equal frequencies, each would be observed 5% of the time. However, some are more common (G, A, L, K) and some rare (C, Y, M, W).

# Normalized frequencies of amino acids: we need these values to calculate denominator p<sub>i</sub>p<sub>i</sub>

Gly	8.9%	Arg	4.1%
Ala	8.7%	Asn	4.0%
Leu	8.5%	Phe	4.0%
Lys	8.1%	Gln	3.8%
Ser	7.0%	lle	3.7%
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 in the genetic code
- These frequencies f<sub>i</sub> sum to I

## Dayhoff model step 3: amino acid substitutions

	Α	R	N	D	С	Q	Е	G	Н	I	L	K	M	F	P	S	T	W	Y	V
	Ala	Arg	Asn	Asp	Cys	Gln	Glu	Gly	His	Ile	Leu	Lys	Met	Phe	Pro	Ser	Thr	Trp	Tyr	Val
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													
Н	21	103	226	43	10	243	23	10												
I	66	30	36	13	17	8	35	0	3											
L	95	17	37	0	У	75	15	17	40	253										
K	57	477	322	85	0	147	104	60	23	43	39									
M	29	17	0	0	0	20	7	7	0	57	207	90								
F	20	7	7	0	0	0	0	17	20	90	167	0	17							
P	345	67	27	10	10	93	40	49	50	7	43	43	4	7						
S	772	137	432	98	117	47	86	450	26	20	32	168	20	40	269					
T	590	20	169	57	10	37	31	50	14	129	52	200	28	10	73	696				
W	0	27	3	0	0	0	0	0	3	0	13	0	0	10	0	17	0			
Y	20	3	36	0	30	0	10	0	40	13	23	10	0	260	0	22	23	6		
V	365	20	13	17	33	27	37	97	30	661	303	17	77	10	50	43	186	0	17	
	Α	R	N	D	С	Q	Е	G	Н	I	L	K	M	F	P	S	T	W	Y	V
	Ala	Arg	Asn	Asp	Cys	Gln	Glu	Gly	His	Ile	Leu	Lys	Met	Phe	Pro	Ser	Thr	Trp	Tyr	Val

From a survey of 1572 observed substitutions, the original amino acid (columns) are compared to the changes (rows).

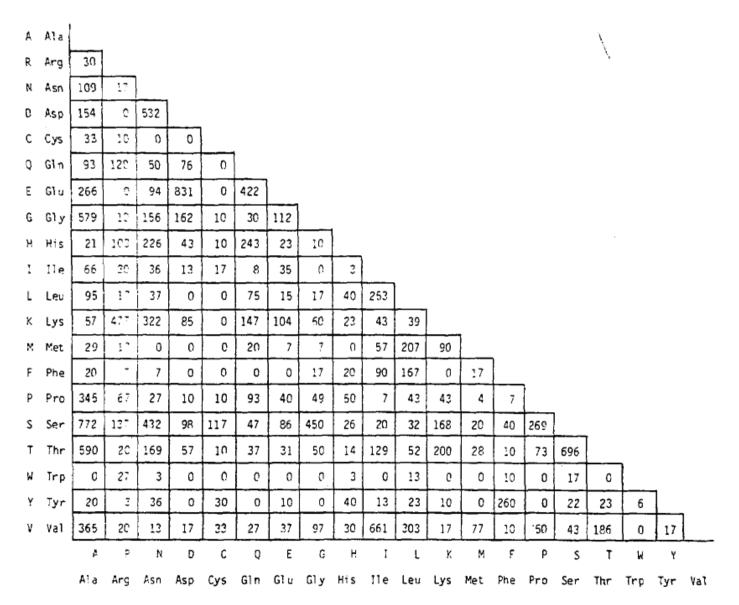


Figure 80. Numbers of accepted point mutations (X 10) accumulated from closely related sequences. Fifteen hundred and seventy-

two exchanges are shown. Fractional exchanges result when ancestral sequences are ambiguous.

### Dayhoff model step 3: amino acid substitutions

	Α	R	N	D	С	Q	Е	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	
Н	21	103	226	43	10	243	23	10

Zooming in on the previous table, note that substitutions are very common (e.g. D  $\rightarrow$  E, A  $\rightarrow$  G) while others are rare (e.g. C  $\rightarrow$  Q, C  $\rightarrow$  E). The scoring system we use for pairwise alignments should reflect these trends.

### Dayhoff model step 3: Relative mutability of amino acids

TABLE 3.2 Relative mutabilities of amino acids. The value of alanine is arbitrarily set to 100.

Asn	134	His	66
Ser	120	Arg	65
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

Dayhoff et al. used (1) data on the frequency of amino acids and (2) data on observed and inferred numbers of substitutions to determine the relative mutability of amino acids. In a scoring system alignment of two tryptophans will be weighted more heavily than two asparagines.

# Dayhoff step 4 (of 7): Mutation probability matrix for the evolutionary distance of 1 PAM

		Original amino acid																			
		A	R	N	D	С	Q	E	G	H	I	L	K	M	F	P	S	T	W	Y	V
_	-	Ala	Arg	Asn	Asp	Cys	Gln	Glu	Gly	His	lle	Leu	Lys	Met	Phe	Pro	Ser	Thr	Trp	Tyr	Val
	A	98.7	0.0	0.1	0.1	0.0	0.1	0.2	0.2	0.0	0.1	0.0	0.0	0.1	0.0	0.2	0.4	0.3	0.0	0.0	0.2
	R	0.0	99.1	0.0	0.0	0.0	0.1	0.0	0.0	0.1	0.0	0.0	0.2	0.0	0.0	0.0	0.1	0.0	0.1	0.0	0.0
	N	0.0	0.0	98.2	0.4	0.0	0.0	0.1	0.1	0.2	0.0	0.0	0.1	0.0	0.0	0.0	0.2	0.1	0.0	0.0	0.0
	D	0.1	0.0	0.4	98.6	0.0	0.1	0.5	0.1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.1	0.0	0.0	0.0	0.0
	C	0.0	0.0	0.0	0.0	99.7	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.1	0.0	0.0	0.0	0.0
	Q	0.0	0.1	0.0	0.1	0.0	98.8	0.3	0.0	0.2	0.0	0.0	0.1	0.0	0.0	0.1	0.0	0.0	0.0	0.0	0.0
_	Е	0.1	0.0	0.1	0.6	0.0	0.4	98.7	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
acid	G	0.2	0.0	0.1	0.1	0.0	0.0	0.1	99.4	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.2	0.0	0.0	0.0	0.1
amino	H	0.0	0.1	0.2	0.0	0.0	0.2	0.0	0.0	99.1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0
	I	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	98.7	0.1	0.0	0.2	0.1	0.0	0.0	0.1	0.0	0.0	0.3
l Ha	L	0.0	0.0	0.0	0.0	0.0	0.1	0.0	0.0	0.0	0.2	99.5	0.0	0.5	0.1	0.0	0.0	0.0	0.0	0.0	0.2
acement	K	0.0	0.4	0.3	0.1	0.0	0.1	0.1	0.0	0.0	0.0	0.0	99.3	0.2	0.0	0.0	0.1	0.1	0.0	0.0	0.0
Repla	M	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.1	0.1	0.0	98.7	0.0	0.0	0.0	0.0	0.0	0.0	0.0
ž	F	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.1	0.1	0.0	0.0	99.5	0.0	0.0	0.0	0.0	0.3	0.0
	P	0.1	0.1	0.0	0.0	0.0	0.1	0.0	0.0	0.1	0.0	0.0	0.0	0.0	0.0	99.3	0.1	0.0	0.0	0.0	0.0
	S	0.3	0.1	0.3	0.1	0.1	0.0	0.1	0.2	0.0	0.0	0.0	0.1	0.0	0.0	0.2	98.4	0.4	0.1	0.0	0.0
	T	0.2	0.0	0.1	0.0	0.0	0.0	0.0	0.0	0.0	0.1	0.0	0.1	0.1	0.0	0.1	0.3	98.7	0.0	0.0	0.1
	W	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	99.8	0.0	0.0
	Y	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.2	0.0	0.0	0.0	0.0	99.5	0.0
	v	0.1	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.6	0.1	0.0	0.2	0.0	0.0	0.0	0.1	0.0	0.0	99.0

This mutation probability matrix includes original amino acids (columns) and replacements (rows). The diagonals show that at a distance of 1 PAM most residues remain the same about 99% of the time (see shaded entries). Note how cysteine (C) and tryptophan (W) undergo few substitutions, and asparagine (N) many.

# **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. At an evolutionary interval of PAM1, one change has occurred over a length of 100 amino acids.
- ➤ Other PAM matrices are extrapolated from PAM1. For PAM250, 250 changes have occurred for two proteins over a length of 100 amino acids.
- ➤ All the PAM data come from closely related proteins (>85% amino acid identity).

# Dayhoff step 4 (of 7): Mutation probability matrix for the evolutionary distance of 1 PAM

									_	
		A	R	N	D	C	Q	E	G	H
		Ala	Arg	Asn	Asp	Cys	Gln	Glu	Gly	His
	A	98.7	0.0	0.1	0.1	0.0	0.1	0.2	0.2	0.0
	R	0.0	99.1	0.0	0.0	0.0	0.1	0.0	0.0	0.1
	N	0.0	0.0	98.2	0.4	0.0	0.0	0.1	0.1	0.2
	D	0.1	0.0	0.4	98.6	0.0	0.1	0.5	0.1	0.0
	C	0.0	0.0	0.0	0.0	99.7	0.0	0.0	0.0	0.0
	Q	0.0	0.1	0.0	0.1	0.0	98.8	0.3	0.0	0.2
	Е	0.1	0.0	0.1	0.6	0.0	0.4	98.7	0.0	0.0
minoacid	G	0.2	0.0	0.1	0.1	0.0	0.0	0.1	99.4	0.0
ino	H	0.0	0.1	0.2	0.0	0.0	0.2	0.0	0.0	99.1

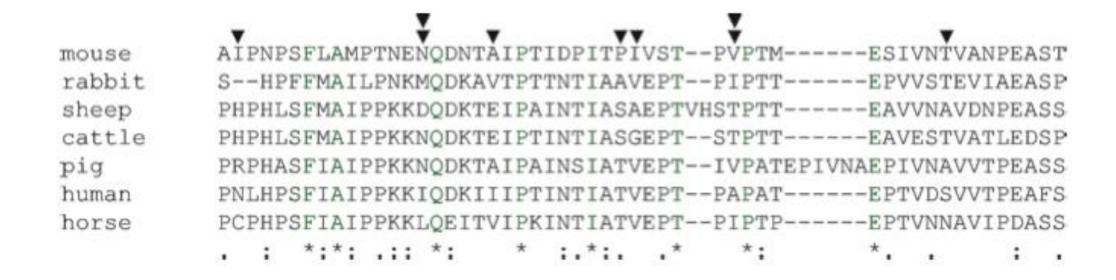
At this evolutionary distance of 1 PAM, 1% of the amino acids have diverged between each pair of sequences. The columns are percentages that sum to 100%.

# Dayhoff step 5 (of 7): PAM250 and other PAM matrices

NP 002037.2	164	IHDNFGIVEGLMTTVHAITATQKTVDGPSGKLWRDGRGALQNII	207
XP 001162057.1	164	IHDNFGIVEGLMTTVHAITATQKTVDGPSGKLWRDGRGALQNII	207
NP 001003142.1	162	IHDHFGIVEGLMTTVHAITATQKTVDGPSGKMWRDGRGAAQNII	205
XP 893121.1	168	IHDNFGIMEGLMTTVHAITATQKTVDGPSGKLWRDGRGAAQNII	211
XP 576394.1	162	IHDNFGIVEGLMTTVHAITATQKTVDGPSGKLWRDGRGAAQNII	205
NP 058704.1	162	IHDNFGIVEGLMTTVHAITATQKTVDGPSGKLWRDGRGAAQNII	205
XP 001070653.1	162	IHDNFGIVEGLMTTVHAITATQKTVDGPSGKLWRDGRGAAQNII	205
XP 001062726.1	162	IHDNFGIVEGLMTTVHAITATQKTVDGPSGKLWRDGRGAAQNII	205
NP 989636.1	162	IHDNFGIVEGLMTTVHAITATQKTVDGPSGKLWRDGRGAAQNII	205
NP 525091.1	161	INDNFEIVEGLMTTVHATTATQKTVDGPSGKLWRDGRGAAQNII	204
XP 318655.2	161	INDNFGILEGLMTTVHATTATQKTVDGPSGKLWRDGRGAAQNII	204
NP 508535.1	170	INDNFGIIEGLMTTVHAVTATQKTVDGPSGKLWRDGRGAGQNII	213
NP 595236.1	164	INDTFGIEEGLMTTVHATTATQKTVDGPSKKDWRGGRGASANII	207
NP 011708.1	162	INDAFGIEEGLMTTVHSLTATQKTVDGPSHKDWRGGRTASGNII	205
XP 456022.1	161	INDEFGIDEALMTTVHSITATQKTVDGPSHKDWRGGRTASGNII	204
NP 001060897.1	166	IHDNFGIIEGLMTTVHAITATQKTVDGPSSKDWRGGRAASFNII	209

Consider a multiple alignment of glyceraldehyde 3-phosphate protein sequences. Some substitutions are observed in columns (arrowheads). These give us insight into changes tolerated by natural selection.

## Dayhoff step 5 (of 7): PAM250 and other PAM matrices



Now consider the alignment of distantly related kappa caseins. There are few conserved column positions, and many some columns (double arrowheads) have five different residues among the 7 proteins.

We want to design a scoring system that is tolerant of distantly related proteins: if the scoring system is too strict then the divergent sequences may be penalized so heavily that authentic homologs are not identified or aligned.

## Dayhoff step 5 (of 7): PAM250 and other PAM matrices

#### original amino acid

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PAM0	Α	R	N	D	С	Q	E	G
Α	100	0	0	0	0	0	0	0
R	0	100	0	0	0	0	0	0
N	0	0	100	0	0	0	0	0
D	0	0	0	100	0	0	0	0
С	0	0	0	0	100	0	0	0
Q	0	0	0	0	0	100	0	0
E	0	0	0	0	0	0	100	0
G	0	0	0	0	0	0	0	100

#### original amino acid

PAM∞	Α	R	N	D	С	Q	Е	G
Α	8.7	8.7	8.7	8.7	8.7	8.7	8.7	8.7
R	4.1	4.1	4.1	4.1	4.1	4.1	4.1	4.1
N	4.0	4.0	4.0	4.0	4.0	4.0	4.0	4.0
D	4.7	4.7	4.7	4.7	4.7	4.7	4.7	4.7
С	3.3	3.3	3.3	3.3	3.3	3.3	3.3	3.3
Q	3.8	3.8	3.8	3.8	3.8	3.8	3.8	3.8
E	5.0	5.0	5.0	5.0	5.0	5.0	5.0	5.0
G	8.9	8.9	8.9	8.9	8.9	8.9	8.9	8.9

At the extreme of perfectly conserved proteins (PAM0) there are no amino acid replacements. At the extreme of completely diverged proteins (PAM∞) the matrix converges on the background frequencies of the amino acids.

## PAM250 matrix: for proteins that share ~20% identity

			Original amino acid																		
		Α														V					
	Α	13	6	9	9	5	8	9	12	6	8	6	7	7	4	11	11	11	2	4	9
	R	3	17	4	3	2	5	3	2	6	3	2	9	4	1	4	4	3	7	2	2
	N	4	4	6	7	2	5	6	4	6	3	2	5	3	2	4	5	4	2	3	3
	D	5	4	8	11	1	7	10	5	6	3	2	5	3	1	4	5	5	1	2	3
	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
acid	Е	5	4	7	11	1	9	12	5	6	3	2	5	3	1	4	5	5	1	2	3
0 9	G	12	5	10	10	4	7	9	27	5	5	4	6	5	3	8	11	9	2	3	7
amino	Н	2	5	5	4	2	7	4	2	15	2	2	3	2	2	3	3	2	2	3	2
	Ι	3	2	2	2	2	2	2	2	2	10	6	2	6	5	2	3	4	1	3	9
en	L	6	4	4	3	2	6	4	3	5	15	34	4	20	13	5	4	6	6	7	13
Replacement	K	6	18	10	8	2	10	8	5	8	5	4	24	9	2	6	8	8	4	3	5
lac	M	1	1	1	1	0	1	1	1	1	2	3	2	6	2	1	1	1	1	1	2
Şeb	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	2	4
	S	9	6	8	7	7	6	7	9	6	5	4	7	5	3	9	10	9	4	4	6
	Т	8	5	6	6	4	5	5	6	4	6	4	6	5	3	6	8	11	2	3	6
	W	0	2	0	0	0	0	0	0	1	0	1	0	0	1	0	1	0	55	1	0
	Y	1	1	2	1	3	1	1	1	3	2	2	1	2	15	1	2	2	3	31	2
	V	7	4	4	4	4	4	4	5	4	15	10	4	10	5	5	5	7	2	4	17

Compare this to a PAM1 matrix, and note the diagonal still has high scores but much information content is lost.

Table 23
Correspondence between Observed Differences and the Evolutionary Distance

Observed Percent Difference	Evolutionary Distance in PAMs				
1	1				
5	5				
10	11				
15	17				
20	23				
25	30				
30	38				
35	47				
40	56				
45	67				
50	80				
55	94				
60	112				
65	133				
70	159				
75	195				
80	246				
85	328				

# Dayhoff step 6 (of 7): from a <u>mutation probability</u> <u>matrix</u> to a relatedness odds matrix

$$R_{ij} = \frac{M_{ij}}{f_i}$$

- 1. A **relatedness odds matrix** reports the probability that amino acid **j** will change to **i** in a homologous sequence.
- 2. The numerator models the observed change. The denominator *fi* is the probability of amino acid residue *i* occurring in the second sequence by chance.
- A positive value indicates a replacement happens more often than expected by chance. A negative value indicates the replacement is not favored.

# 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.
- 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,

$$s_{i,j} = 10 \times \log\left(\frac{q_{i,j}}{p_{i,j}}\right)$$

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

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

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

A score of 0 is neutral.

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.

# Dayhoff step 7 (of 7): log odds scoring matrix

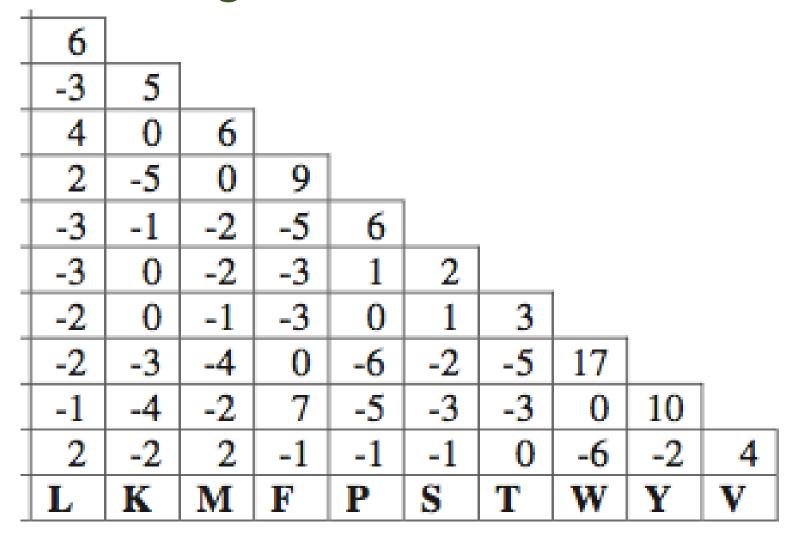
$$s_{ij} = 10 \times \log_{10} \left( \frac{M_{ij}}{f_i} \right).$$

- > A log odds matrix is the logarithmic form of the relatedness odds matrix.
- > <u>sij</u> is the score for aligning any two residues in a pairwise alignment. (There is also a score for aligning a residue with itself.)
- > Mij is of the observed frequency of substitutions for each pair of amino acids. These values ("target frequencies") are derived from a mutation probability matrix.

# Example of a score for aligning cysteine and leucine using the values in a PAM250 scoring matrix

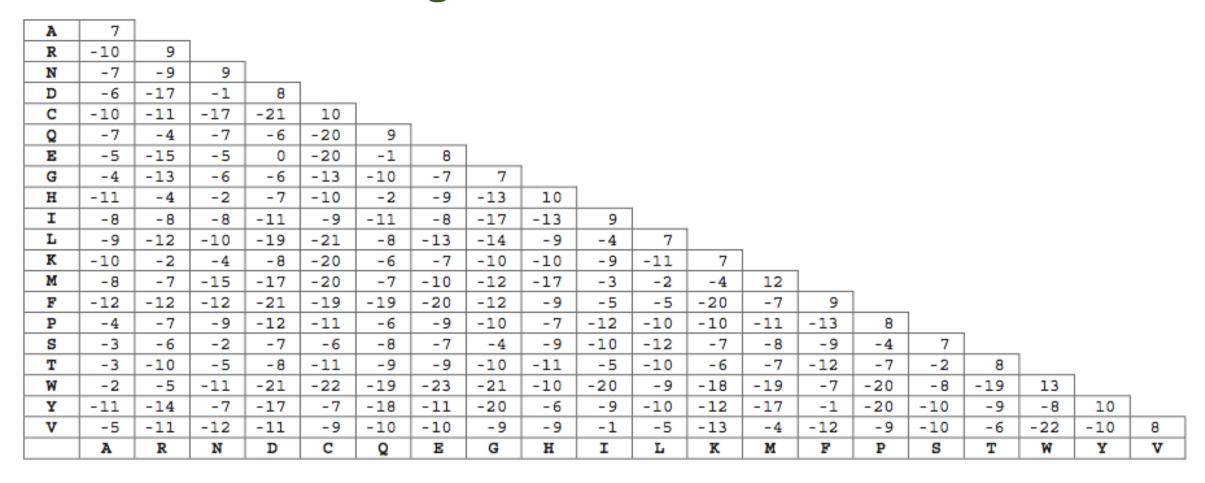
$$s_{\text{(cysteine, leucine)}} = 10 \times \log_{10} \left( \frac{0.02}{0.085} \right) = -6.3$$

# **Log-odds matrix for PAM250**



This is a useful matrix for comparing **distantly related proteins**. Note that an alignment of two tryptophan (W) residues earns +17 and a W to T mismatch is -5.

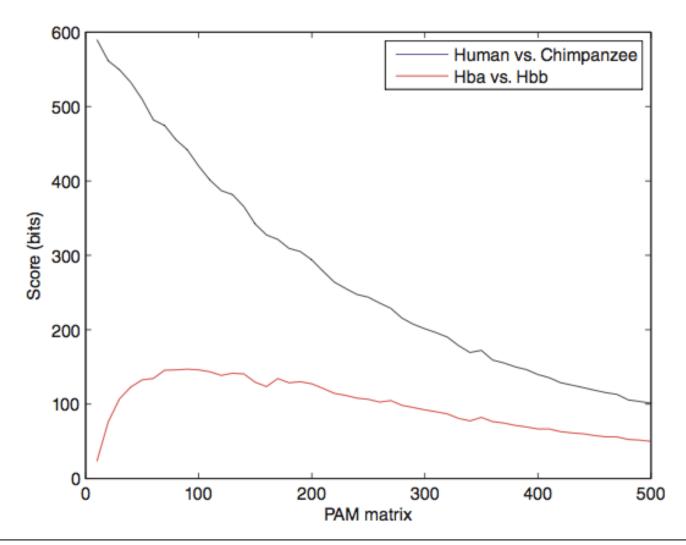
# **Log-odds matrix for PAM10**



#### More closely related proteins

This is an example of a scoring matrix with "severe" penalties. A match of W to W earns +13, but a mismatch (e.g. W aligned to T) has a score of -19, far lower than in PAM250.

### **Effect of scoring matrix on bit scores**



Look at score for distantly related proteins (e.g. beta globin versus alpha globin) and note that PAM10 or similar matrices assign very low scores. This effect is not seen for very closely related proteins (e.g. a chimp vs. human globin).

**PAM matrices** are based on data from the alignment of closely related protein families, and they involve the assumption that substitution probabilities for highly related proteins (e.g., PAM40) can be extrapolated to probabilities for distantly related proteins (e.g., PAM250).

**BLOSUM matrices** are based on empirical observations of more distantly related protein alignments.

## **BLOSUM62** scoring matrix

blocks substitution matrix

A	4						plo	CKS S	subs	titut	ion r	natr	IX							
		5	1																	
R	-1			1																
N	-2	0	6		1															
D	-2	-2	1	6																
C	0	-3	-3	-3	9															
Q	-1	1	0	0	-3	5														
E	-1	0	0	2	-4	2	5													
G	0	-2	0	-1	-3	-2	-2	6												
H	-2	0	1	-1	-3	0	0	-2	8											
I	-1	-3	-3	-3	-1	-3	-3	-4	-3	4										
L	-1	-2	-3	-4	-1	-2	-3	-4	-3	2	4									
K	-1	2	0	-1	-1	1	1	-2	-1	-3	-2	5								
M	-1	-2	-2	-3	-1	0	-2	-3	-2	1	2	-1	5							
F	-2	-3	-3	-3	-2	-3	-3	-3	-1	0	0	-3	0	6						
P	-1	-2	-2	-1	-3	-1	-1	-2	-2	-3	-3	-1	-2	-4	7					
S	1	-1	1	0	-1	0	0	0	-1	-2	-2	0	-1	-2	-1	4				
T	0	-1	0	-1	-1	-1	-1	-2	-2	-1	-1	-1	-1	-2	-1	1	5			
$\mathbf{W}$	-3	-3	-4	-4	-2	-2	-3	-2	-2	-3	-2	-3	-1	1	-4	-3	-2	11		
Y	-2	-2	-2	-3	-2	-1	-2	-3	2	-1	-1	-2	-1	3	-3	-2	-2	2	7	
V	0	-3	-3	-3	-1	-2	-2	-3	-3	3	1	-2	1	-1	-2	-2	0	-3	-1	4
	A	R	N	D	C	Q	E	G	H	I	L	K	M	F	P	S	T	W	Y	V

BL62 is the default scoring matrix at the NCBI BLAST site.

#### Derivation of BLOSUM matrices

BLOSUM matrices are derived from comparisons of blocks of sequences from the Blocks database.

What are blocks and what is the blocks database?

A *block* is an ungapped multiple alignments of highly conserved, short regions. Here is what a sample block looks like:



Conserved blocks in alignment

The blocks database contains multiple alignments of conserved regions in protein families.

https://snipcademy.com/pairwise-alignment#blosum---blocks-substitution-matrix

# **BLOSUM Matrices**

BLOSUM matrices are based on local alignments.

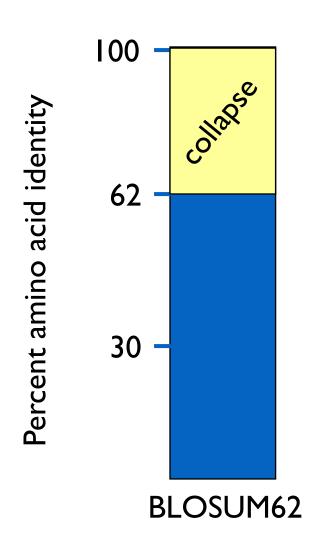
All BLOSUM matrices are based on observed alignments; they are not extrapolated from comparisons of closely related proteins.

BLOSUM stands for blocks substitution matrix.

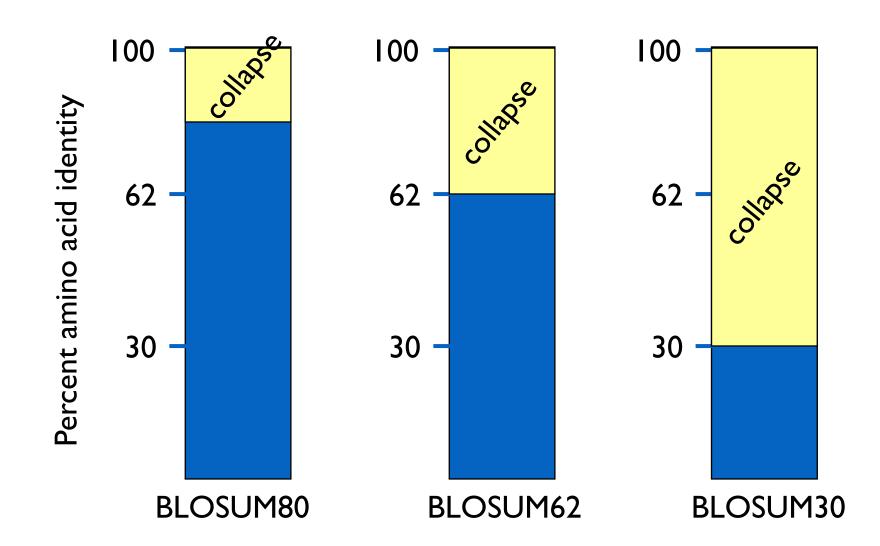
BLOSUM62 is a matrix calculated from comparisons of sequences with no less than 62% divergence.

BLOSUM62 is the default matrix in BLAST 2.0.

# **BLOSUM Matrices**



# **BLOSUM Matrices**

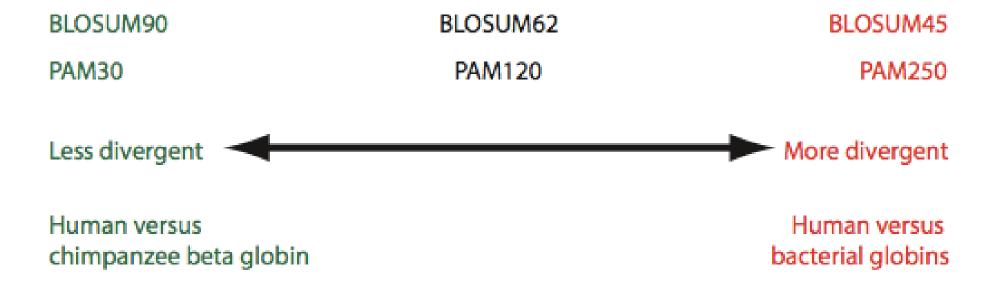


PAM matrices with similar BLOSUM matrices (relative entropy of each PAM matrix is from Altschul 1991)<sup>[18]</sup>

PAM matrix	Equivalent BLOSUM matrix	Relative entropy (bits)
PAM100	Blosum90	1.18
PAM120	Blosum89	0.98
PAM160	Blosum60	0.70
PAM200	Blosum52	0.51
PAM250	Blosum45	0.36

Source: Wikipedia

# Summary of PAM and BLOSUM matrices



A higher PAM number, and a lower BLOSUM number, tends to correspond to a matrix tuned to more divergent proteins.