Lessons 3 & 4. Contents

1. Introduction to proteins

2. Sequence alignments – Part 1

3. Substitutions and gaps

4. Homology search in data banks

BIOinformatics = genes + proteins + informatics (part of computational biology, biocomputing)

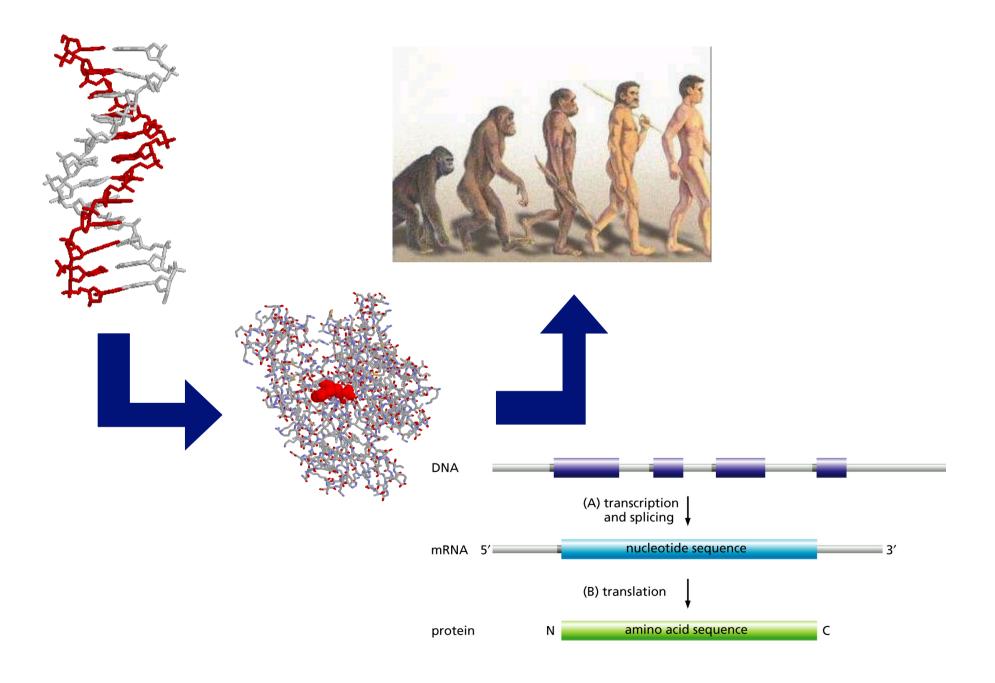
GENE: DNA segment which codes for a specific protein and determines an hereditary feature

Building blocks: 4 nucleotides (ACGT/U)

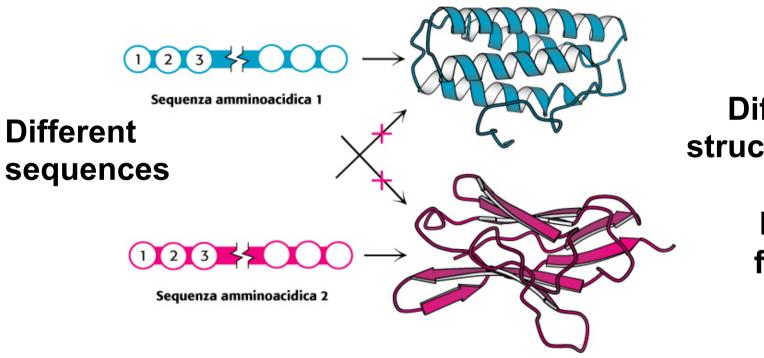
PROTEIN: expression product of a gene and ed EFFECTOR of the biochemical function whose information is stored in the gene

Building blocks: 20 amino acids

MOLECULAR EVOLUTION



Proteins are made up of 20 different amino acids: ACDEFGHIKLMNPQRSTVWY



Different 3D structures and different biological functions!

Knowing the <u>relationship between a protein structure and its</u> <u>function</u> provides a fundamental understanding of how the protein works allowing to foresee how modifying the structure could affect the function

Most of the currently <u>marketed pharmaceuticals</u> act by interacting with proteins

The structure adopted by a protein is entirely <u>determined by its</u> <u>amino acids sequence</u>, however the rules that govern how a protein chain of a given sequence folds up are not yet fully understood

One of the main aims of Bioinformatics is to predict and analyze the structure of proteins and the relationship of the structure to the function

Proteins are made of 20 amino acids, covalently bonded by peptide bonds

The 20 amino acids are made of C, N, O, H (S in case of Cys and Met) atoms

Their side chains differ in size and chemical nature

NONPOLAR SIDE CHAINS



(Ala, or A)

valine

(Val, or V)

leucine

(Leu, or L)

isoleucine

(Ile, or I)

proline

(Pro, or P)

phenylalanine

(Phe, or F)

methionine

(Met, or M)

tryptophan

(Trp, or W)

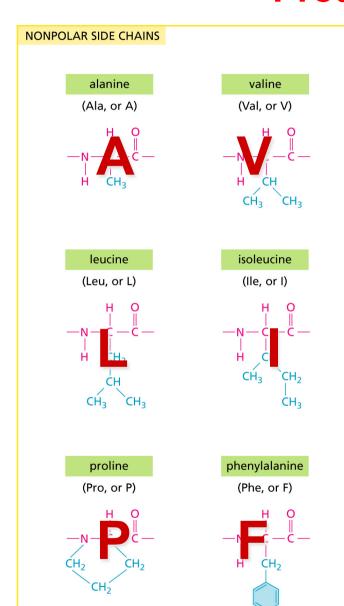
glycine

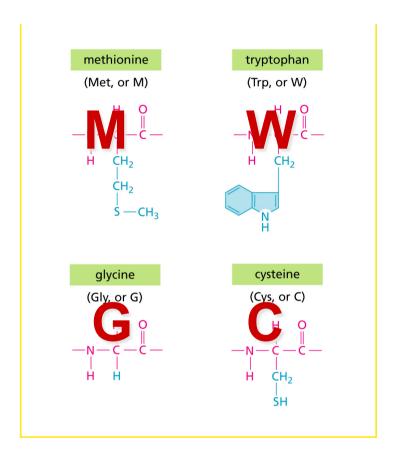
(Gly, or G)

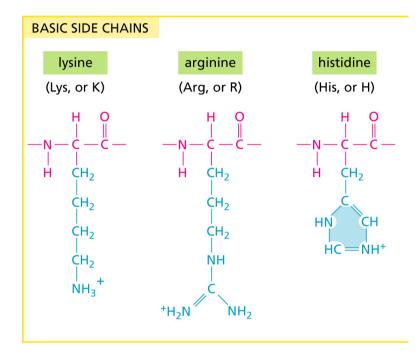
cysteine

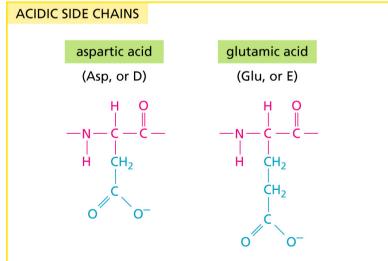
(Cys, or C)

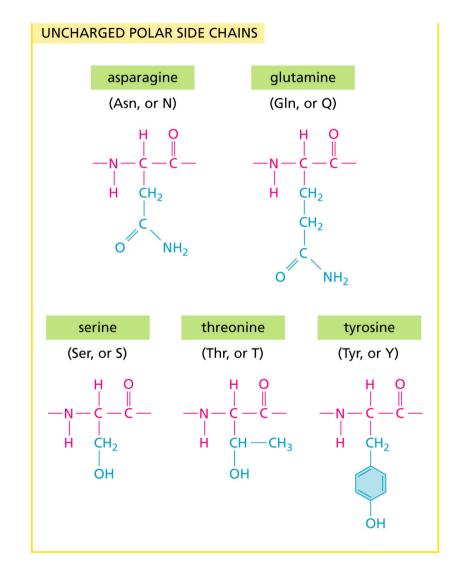
$$\begin{array}{c|c} & H & O \\ & | & | \\ -N - C - C - C - \\ & | & | \\ H & CH_2 \\ & | \\ SH \end{array}$$

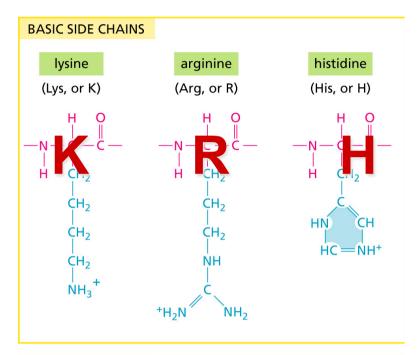


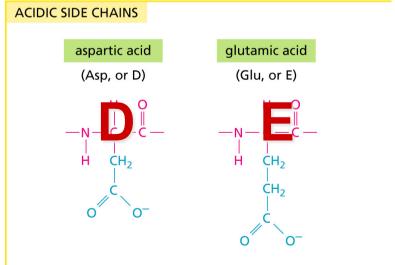


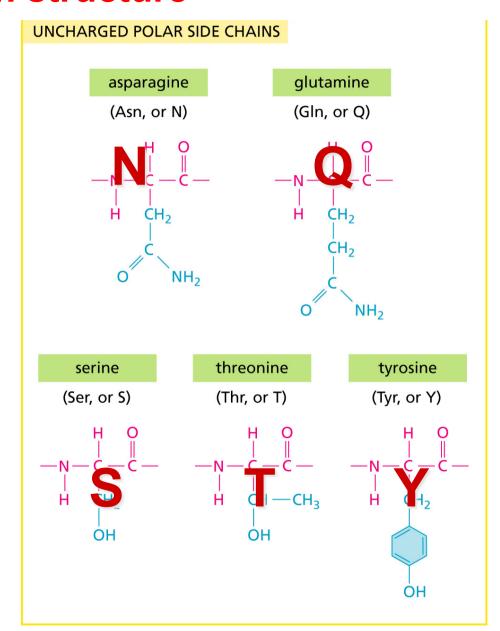




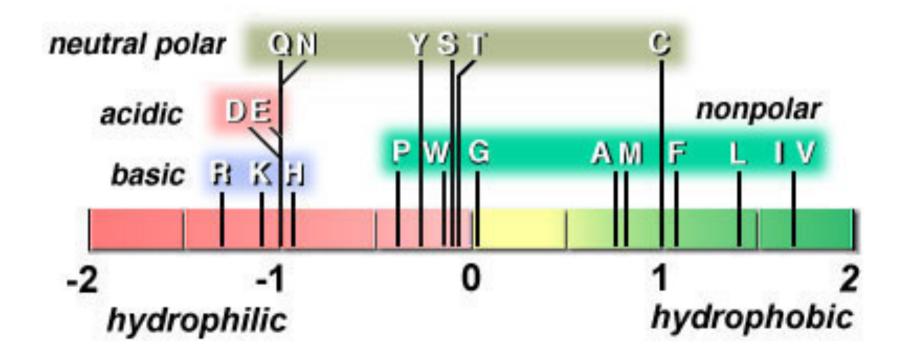




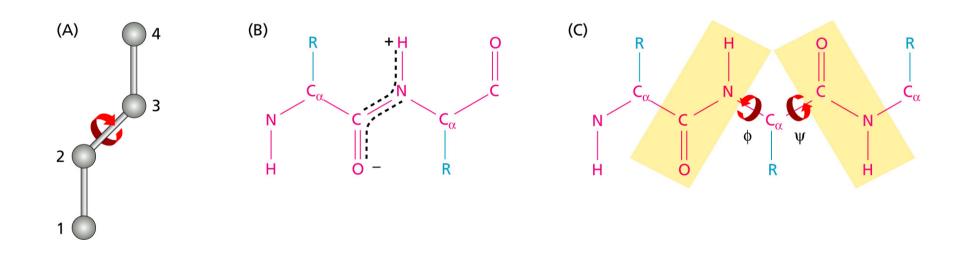




Amino acids hydrophilicity/hydrophobicity:

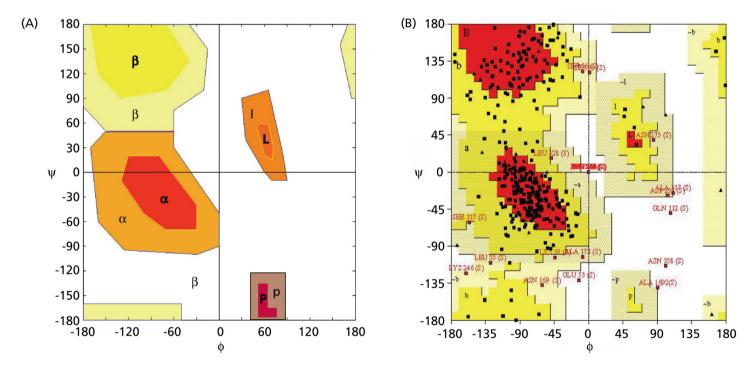


Peptide bonds are planar. However the bonds made by $C\alpha$ with N and C are singular and give rise to two torsional angles per residue (ϕ and ψ , defined between -180° and +180°)



These torsion angles are the main source of flexibility for proteins

 ϕ and ψ angles assume preferentially some values, as steric hindrance prevents certain combinations



Ramachandran plot, the darker the color, the more favorable the combination of angles

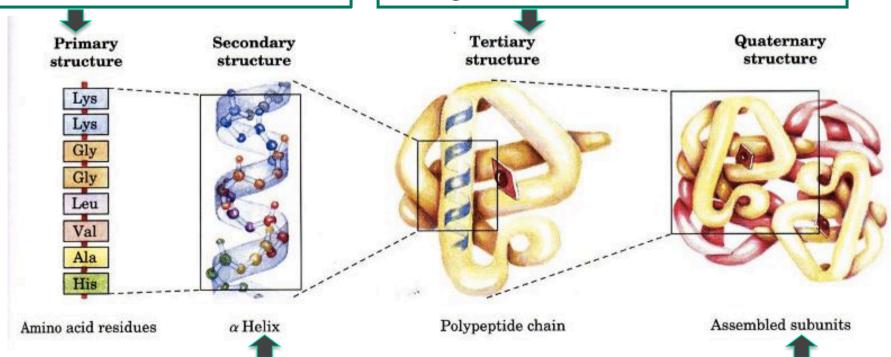
Example for a real structure, with outliers

Protein structure: hierarchy

There are four levels of protein structure to consider:

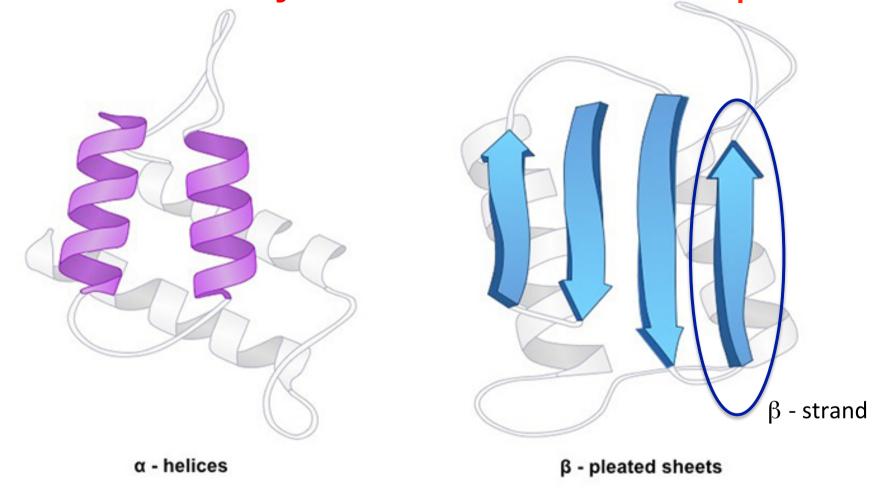
The protein sequence: types and order of the amino acids

Folding and packing of the secondary structure elements to give the final 3D structure



1st level of folding where parts of the protein fold to form local repetitive structures Many functional proteins are formed by more protein chains (identical or not)

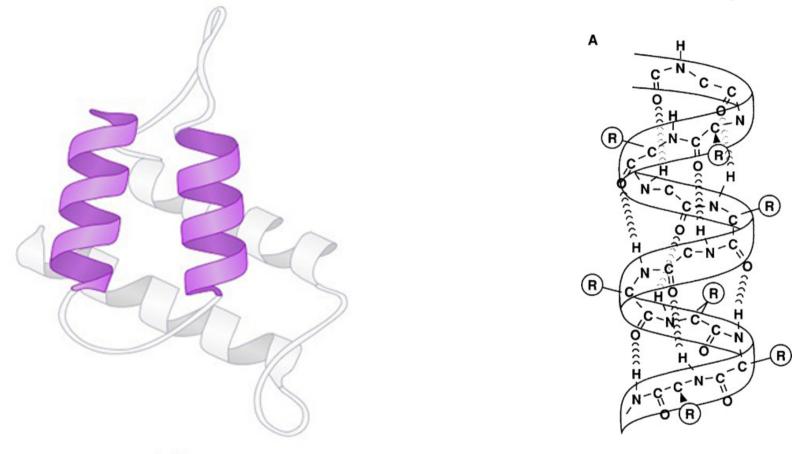
Protein secondary structure: α helices and β -strands



 α -helices and β -strands are the only regular protein secondary structure motifs

 α -helices and β -strands are connected by turns (ordered 3/4-residue motifs) or loops

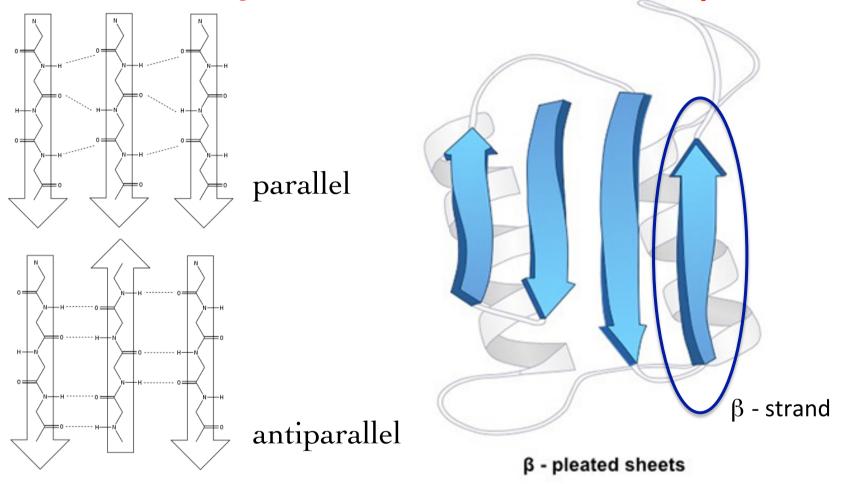
Protein secondary structure: α helices and β -strands



α - helices

In a α -helix (right hand) conformation dihedral angles (ϕ , ψ) assume values around $-60^{\circ}/-45^{\circ}$ and every backbone N-H group hydrogen bonds to the backbone C=O group of the amino acid located 4-residue upstream

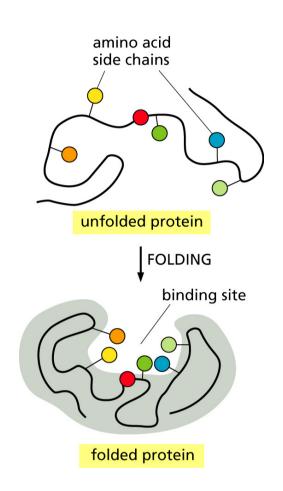
Protein secondary structure: α helices and β -strands



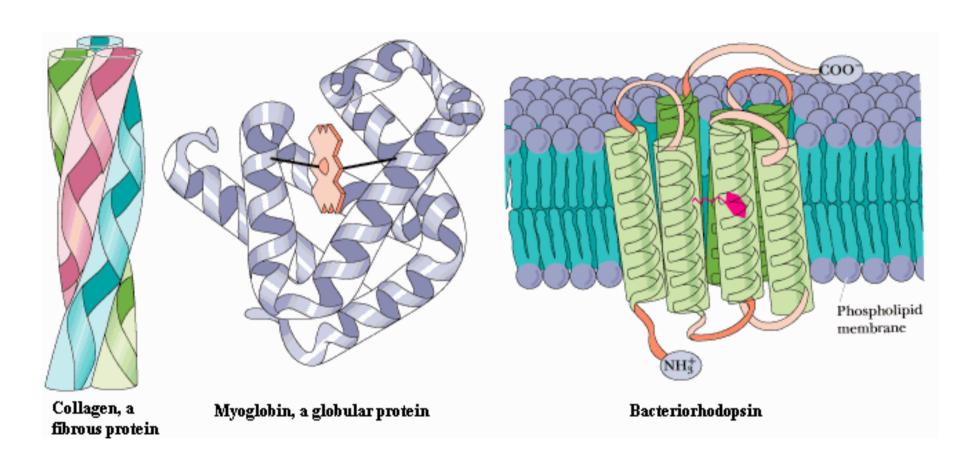
A β-strand is a stretch of polypeptide chain typically 3 to 10 amino acids long with backbone in an extended conformation - dihedral angles (φ, ψ) values around $-135^{\circ}/+135^{\circ}$; they can form sheets where their backbone H-bond to that of another strand

The folded state of a protein corresponds to a free energy minimum

Residues which are distant in sequence can come close in the folded structure to form a functional site, e.g. a binding/catalytic site



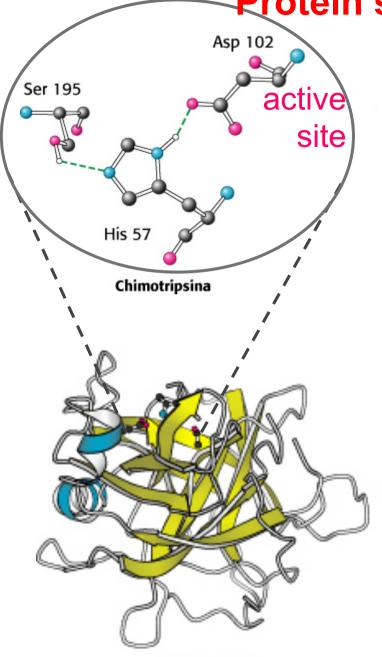
Protein structure: classes



structural

globular

membrane



Chimotripsina

Enzymes are globular proteins which catalyze reactions through an active site

Example:

Chymotrypsin is a digestive enzyme active in the small intestine where it contributes to proteins deigestion

It can break peptide bonds thanks to its active site (catalytic Ser-protease triad). **Protein domains**

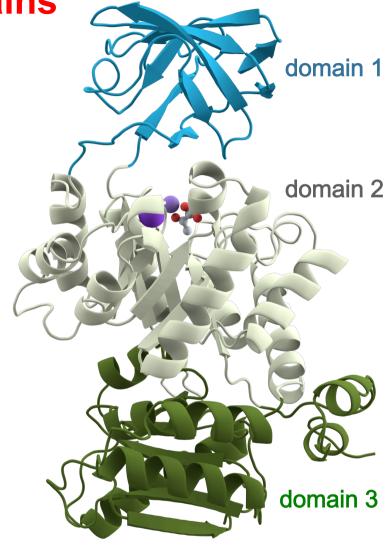
Many proteins consist of several domains

A protein domain is a region of a protein that is self-stabilizing and that <u>folds independently</u> from the rest

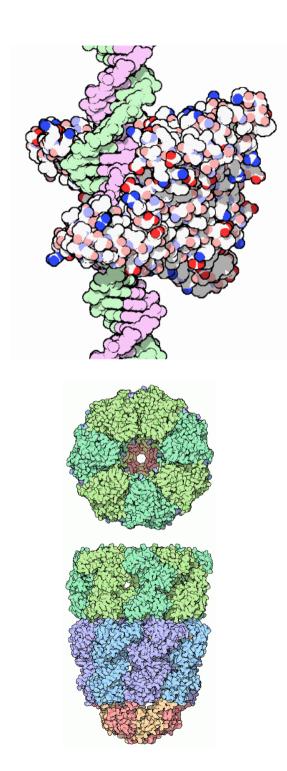
Domains usually form <u>functional units</u>

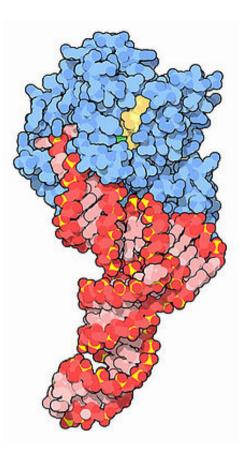
Domains vary in length from ≈50 amino acids up to ≈250 amino acids

Molecular evolution uses domains as building blocks: a domain may appear in a variety of different proteins

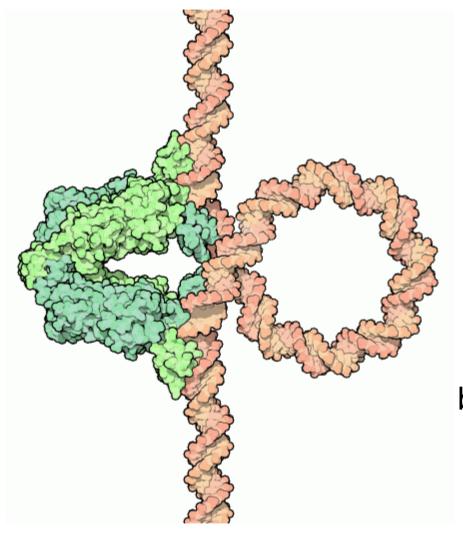


Example: pyruvate kinase contains an all- β nucleotide binding domain (blue), an α/β -substrate binding domain (grey) and an α/β -regulatory domain (green) connected by linkers





A specific protein fold allows it to perform its function!



lac repressor:

blocks transcription of a specific DNA region

Example of protein sequence

YQVRNSSGLYHVTNDCPNSSIVYEAADAILHTPGCVPCVREGNASRCWV
AVTPTVATRDGKLPTTQLRRHIDLLVGSATLCSALYVGDLCGSVFLVGQ
LFTFSPRRHWTTQDCNCSIYPGHITGHRMAWDMMMNWSPTAALVVAQLL
RIPQAILDMIAGAHWGVLAGIAYFSMVGNWAKVLVVLLLFAGVDAETHV
TGGSAGHTTAGLVRLLSPGAKQNIQLINTNGSWHINSTALNCNESLNTG
WLAGLFYHHKFNSSGCPERLASCRRLTDFAQGGGPISYANGSGLDERPY
CWHYPPRPCGIVPAKSVCGPVYCFTPSPVVVGTTDRSGAPTYSWGANDT
DVFVLNNTRPPLGNWFGCTWMNSTGFTKVCGAPPCVIGGVGNNTLLCPT
DCFRKHPEATYSRCGSGPWITPLLLLLALPQRAY

Structural/functional information is contained in the amino acid sequence of a protein chain

Proteins vary by the different combination of the 20 amino acids in their sequence

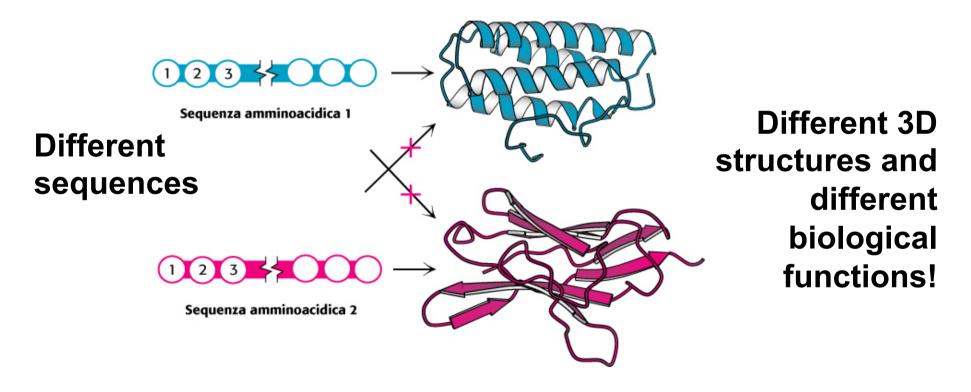
Example of protein sequence

YQVRNSSGLYHVTNDCPNSSIVYEAADAILHTPGCVPCVREGNASRCWV
AVTPTVATRDGKLPTTQLRRHIDLLVGSATLCSALYVGDLCGSVFLVGQ
LFTFSPRRHWTTQDCNCSIYPGHITGHRMAWDMMMNWSPTAALVVAQLL
RIPQAILDMIAGAHWGVLAGIAYFSMVGNWAKVLVVLLLFAGVDAETHV
TGGSAGHTTAGLVRLLSPGAKQNIQLINTNGSWHINSTALNCNESLNTG
WLAGLFYHHKFNSSGCPERLASCRRLTDFAQGGGPISYANGSGLDERPY
CWHYPPRPCGIVPAKSVCGPVYCFTPSPVVVGTTDRSGAPTYSWGANDT
DVFVLNNTRPPLGNWFGCTWMNSTGFTKVCGAPPCVIGGVGNNTLLCPT
DCFRKHPEATYSRCGSGPWITPLLLLLALPQRAY

Structural/functional information is contained in the amino acid sequence of a protein chain

In principle, there can be 20^n different polypeptide chains of length n: 20^{250} of length 250 (over 10^{325}), but only a tiny fraction of them exist (again, think of evolution!)

Proteins are made up of 20 different amino acids: ACDEFGHIKLMNPQRSTVWY



There seems to be a limited number, in the order of thousands (10³), of fold families, thus also proteins with different sequences may in principle fold similarly

"Informatics" problems with protein sequences

Storing and archiving protein sequences

Search for regularities and "patterns" (e.g. active sites)

Comparing protein sequences and measuring their similarity

"Informatics" problems with protein sequences

Storing and archiving protein sequences

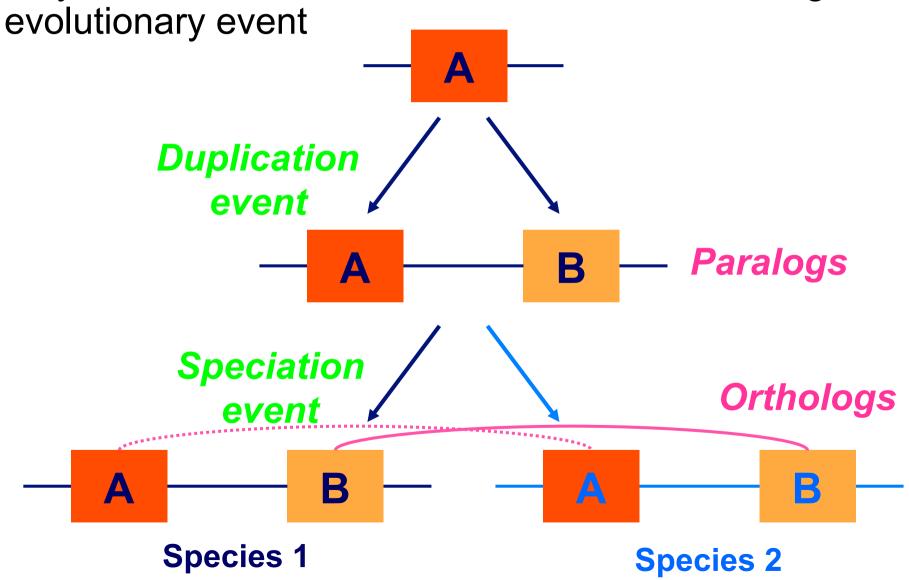
Search for regularities and "patterns" (e.g. active sites)

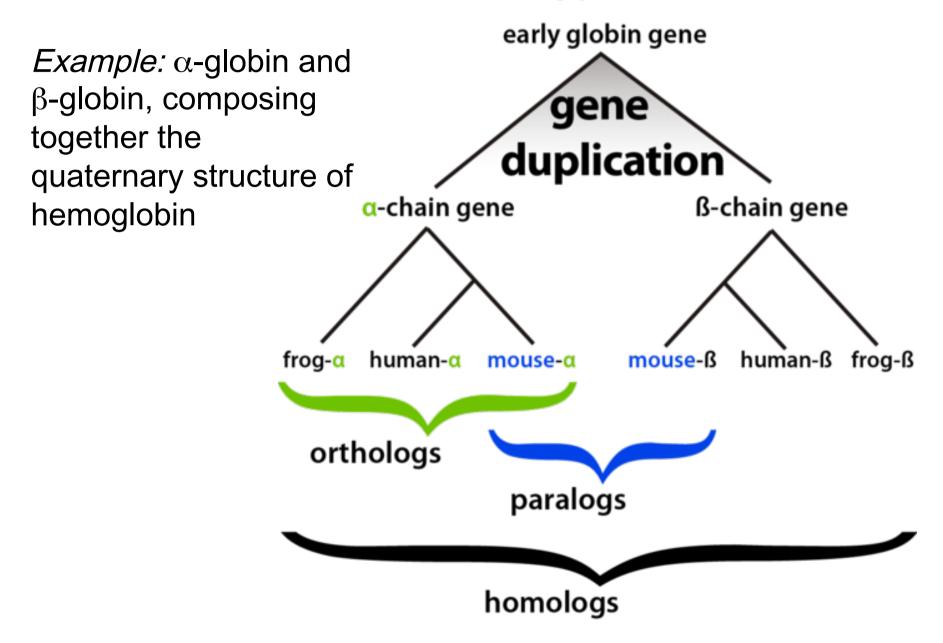
Comparing protein sequences and measuring their similarity

Similarity & homology

- Two sequences are similar if they can be aligned so that many corresponding (aligned) amino acids are identical or similar
- Technically two or more proteins may be defined homologous if they derive from a common ancestor
- Homology between two sequences cannot be observed but only inferred by their similarity in sequence or function
- The concept of similarity can be extended to 3D structures

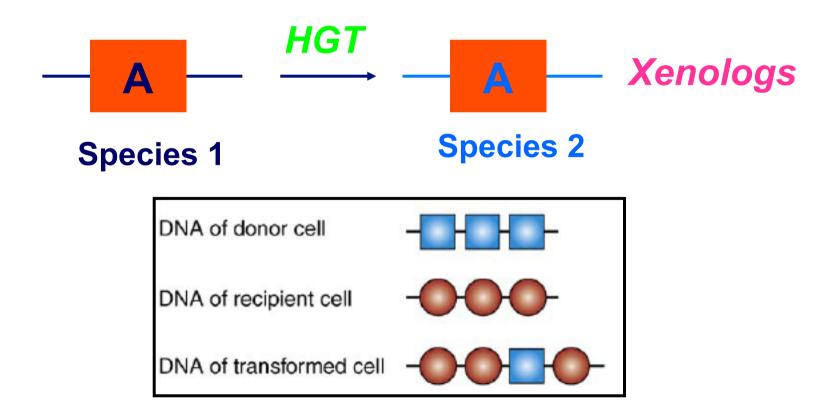
Two or more proteins may be defined homologous if they derive from a common ancestor through an evolutionary event



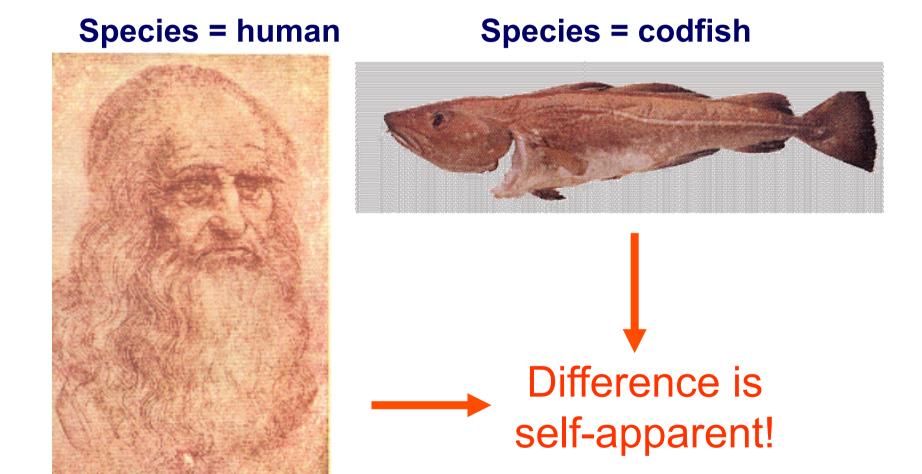


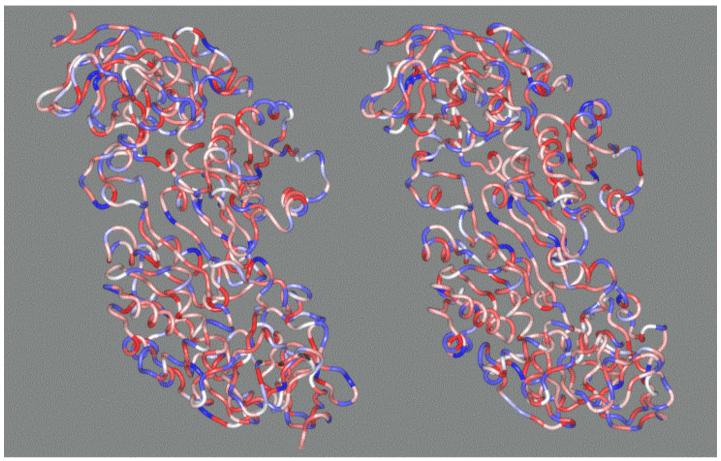
Two sequences are homologous if they derive from a common ancestor:

The Horizontal Gene Transfer (HGT) is the genetic material transfer between two different genomes

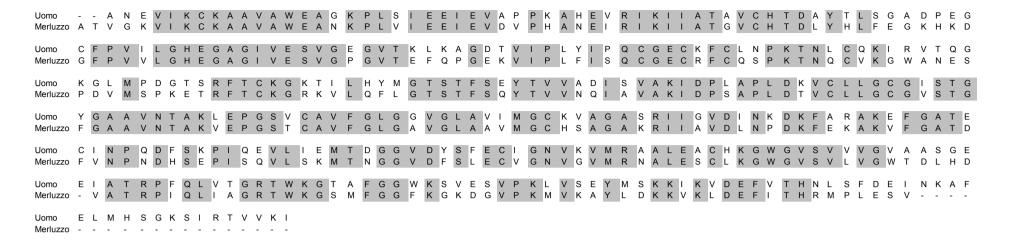


Homology between two proteins/genes can be deduced by their similarity in sequence, structure or function





Species = codfish **Species = human** - - A N E V I K C K A A V A W E A G K P L S I Merluzzo A T V G K V I K C K A A V A W E A N K P L V I E E I E V D V P H A N E I R I K I I A T G V C H T D L Y H L F E G K H K D C F P V I L G H E G A G I V E S V G E G V T K L K A G D T V I P L Y I P Q C G E C K F C L N P K T N L C Q K I R V T Q G Merluzzo G F P V V L G H E G A G I V E S V G P G V T E F Q P G E K V I P L F I S Q C G F C R F C Q S P K T N Q C V K G W A N E S TVVADISAKI G K T I L Y M G T S Merluzzo P D V M S P K E T R F T C K G R K V L TVVNQL F G L G G V G L A V I M G C K V A G A S R I I G V D I N K D K F A R A K E F G A T E Merluzzo F G A A V N T A K V E G L G A V M L A A V M G C S A G A K R I I A V D L N P D K F E K A K V F G A T D IQEVLIEMT N V K V M R A A L E A C H K G W G V S V V V G V A A S G E Merluzzo F V N P N D H S E P I S Q V L S K M T Uomo Merluzzo - - -



Two sequences are similar if they can be *aligned* so that many corresponding (aligned) amino acids (or nucleotides) are identical or similar

With a sequence alignment we search for a correspondence between amino acids (or nucleotides) that most probably reflects the evolution of proteins (or genes)

Sequence alignment

What is the correspondence between amino acids (or nucleotides) which most likely reflects the evolution of two proteins (or genes)?

Aim:

minimizing the evolutionary distance between sequences to be aligned, therefore minimizing differences (that is maximizing similarities) between the components (nucleotides or amino acids) of the sequences themselves

The *hypothesis* of *most reasonable* alignment is the one involving the lowest number of mutations to pass from one sequence to the other one

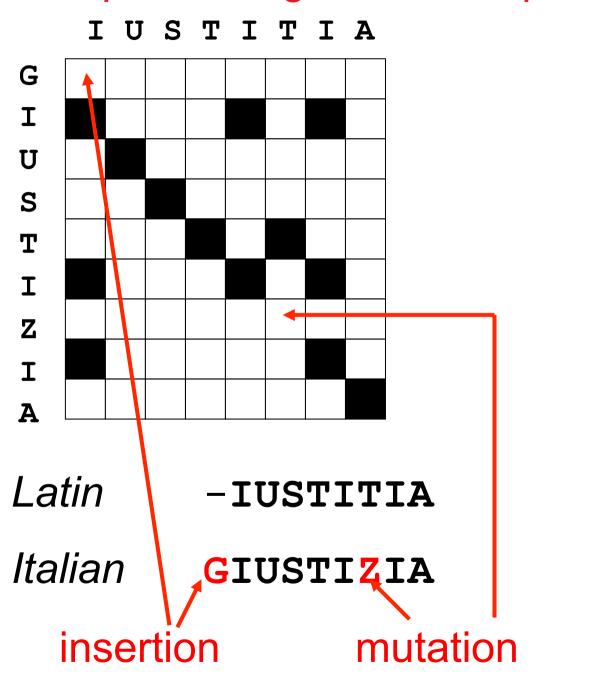
Applications

- Function recognition
 - assessing a significant similarity between two sequences is enough
- Phylogeny
 - Measuring the similarity on a quantitative basis is required
- Model building
 - Explicitly constructing the best possible alignment is required

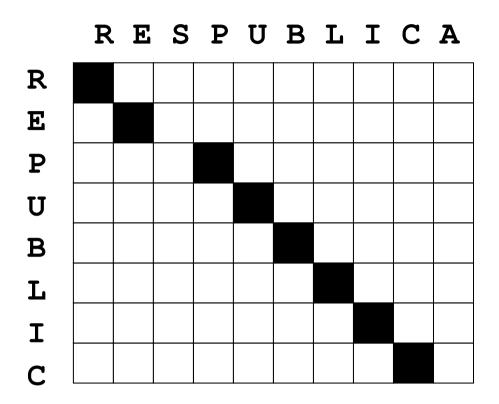
A dot matrix or **dot-plot** provides un immediate view of the similarity between two sequences

In a **dot-plot** a dot is reported in correspondence of two identical characters

Sequence alignment: dot-plot



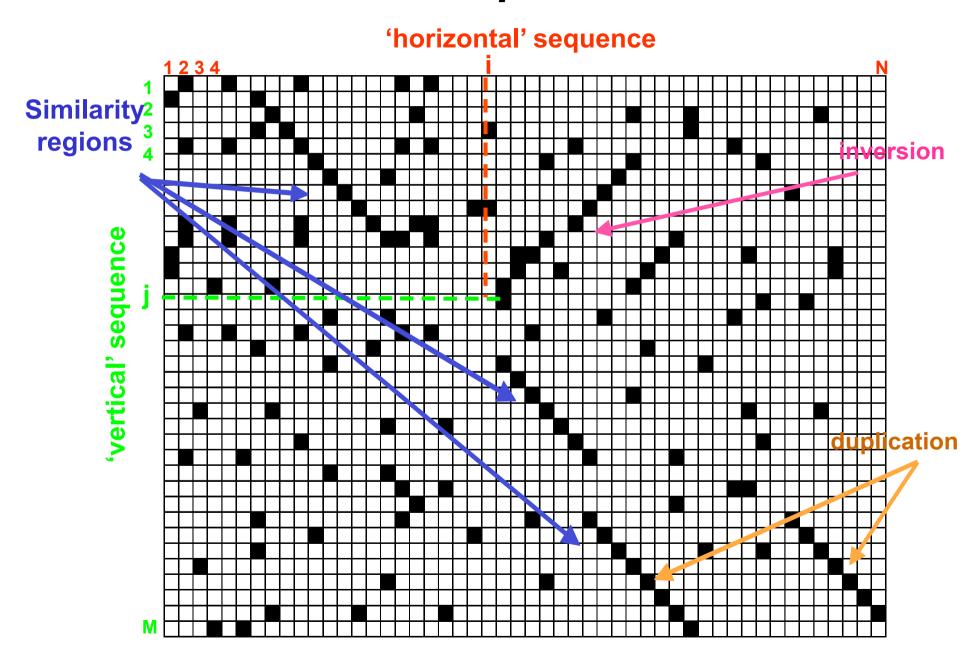
Sequence alignment: dot-plot



Latin RESPUBLICA

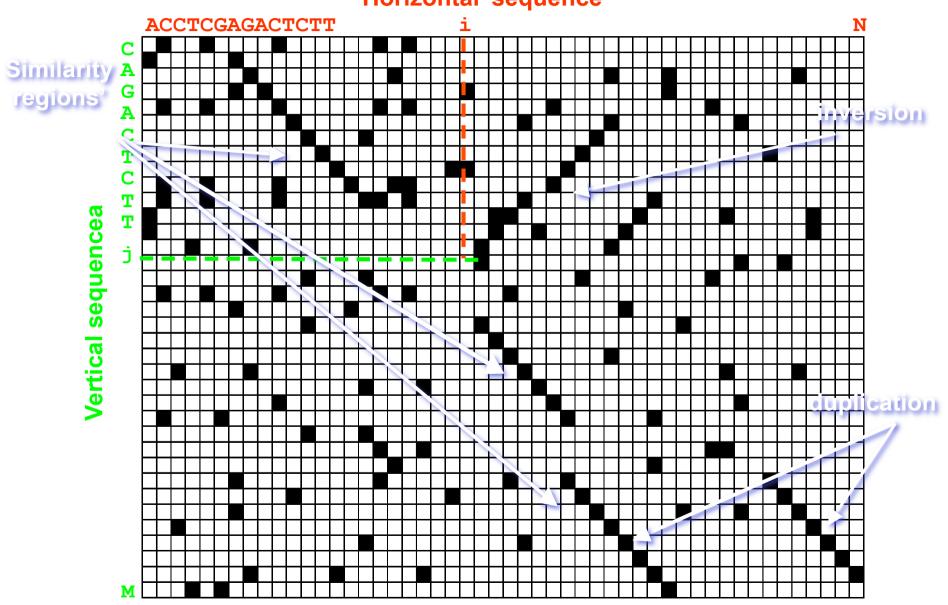
English RE-PUBLIC
deletions

Dot plot



Dot plot

'Horizontal' sequence



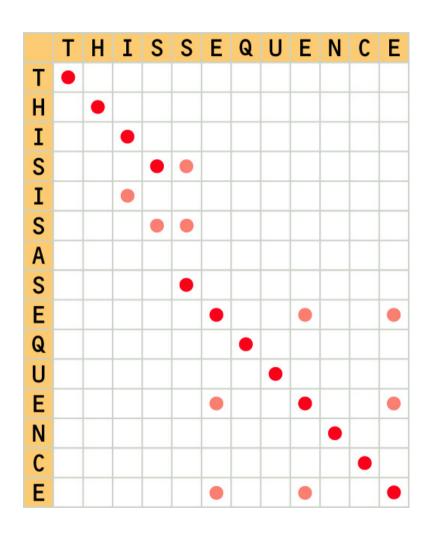
Sequence alignment: dot-plot

Sequence1 (12 charaters)

THISSEQUENCE

Sequence2 (15 charaters)

THISISASEQUENCE



Sequence alignment: dot-plot

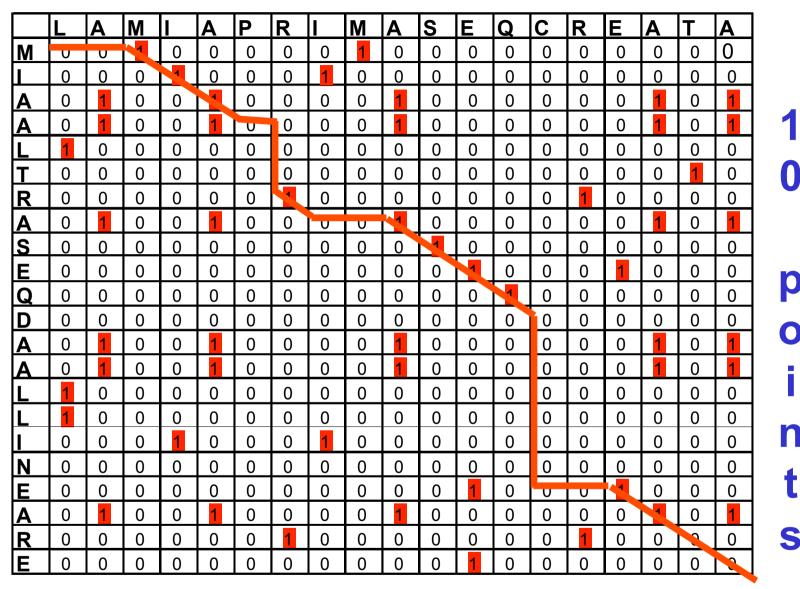
Sequence1 (19 characters)

LAMIAPRIMASEQCREATA

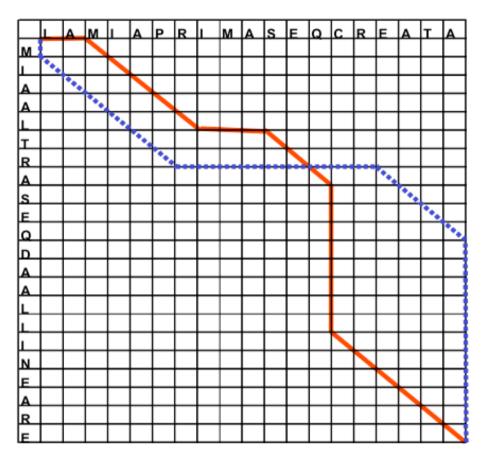
Sequence2 (22 characters)

MIAALTRASEQDALLINEARE

	L	Α	M	I	Α	Р	R	I	M	Α	S	Е	Q	С	R	Ε	Α	Т	Α
M	0	0	1	0	0	0	0	0	1	0	0	0	0	0	0	0	0	0	0
	0	0	0	1	0	0	0	1	0	0	0	0	0	0	0	0	0	0	0
Α	0	1	0	0	1	0	0	0	0	1	0	0	0	0	0	0	1	0	1
Α	0	1	0	0	1	0	0	0	0	1	0	0	0	0	0	0	1	0	1
L	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
T	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0
R	0	0	0	0	0	0	1	0	0	0	0	0	0	0	1	0	0	0	0
Α	0	1	0	0	1	0	0	0	0	1	0	0	0	0	0	0	1	0	1
S	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0	0
E	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	1	0	0	0
Q	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0
D	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
Α	0	1	0	0	1	0	0	0	0	1	0	0	0	0	0	0	1	0	1
Α	0	1	0	0	1	0	0	0	0	1	0	0	0	0	0	0	1	0	1
L	1	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
L	1	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	1	0	0	0	0	0	0	0	0	0	0	0
N	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
E	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	1	0	0	0
Α	0	1	0	0	1	0	0	0	0	1	0	0	0	0	0	0	1	0	1
R	0	0	0	0	0	0	1	0	0	0	0	0	0	0	1	0	0	0	0
Е	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0	0	0	0	0



LAMIAP---RIMASEQ-----CREATA
--MIA-ALTR--ASEQDAALLIN--EARE



LAMIAPRIMASEQCREATA-----MIAALTR-----ASEQDAALLINEARE
LAMIAPRIMASEQ-----CREATA
--MIAAL---TRASEQDAALLINEARE

Different paths through a dot-plot correspond to different alignments

The quality of an alignment is measured by giving it a quantitative score

Non only the identity between amino acids matter, but also the similarity

Not all amino acids substitutions are equally likely to occur

For obtaining an optimal alignment (the one with maximum score, not necessarily the correct one, reflecting the evolutionary process), we need:

- A score for the substitution of amino acids/ nucleobases
- 2) Penalty for insertions/deletions (INDELs)
- 3) Algorithm to perform the alignment
- 4) Measure of the alignment significance

Sequence alignment – Part 1

For obtaining an optimal alignment (the one with maximum score, not necessarily the correct one, reflecting the evolutionary process), we need:

- A score for the substitution of amino acids/ nucleobases
- 2) Penalty for insertions/deletions (INDELs)
- 3) Algorithm to perform the alignment
- 4) Measure of the alignment significance

1. Score for substitutions

- Identity (1 or 0) between nucleobases and amino acids
- Physico-chemical properties of amino acids
- Lowest number of nucleobases to be substituted to obtain the observed mutation
- Substitution frequences observed in protein families (first proposed by Margaret Dayhoff in the '70s)



G A V L I

S C T M P

F Y W H K R

D E N Q

	A	C	D	Е	F	G	Н	I	K	L	M	N	P	Q	R	S	T	V	W	Y
A	2																			
C	-2	12																		
D	0	-5	4																	
E	0	-5	3	4																
F	-4	-4	-6	-5	9															
G	1	-3	1	0	-5	5														
Н	-1	-3	1	1	-2	-2	6													
I	-1	-2	-2	-2	1	-3	-2	5												
K	-1	-5	0	0	-5	-2	0	-2	5											
L	-2	-6	-4	-3	2	-4	-2	2	-3	6										
M	-1	-5	-3	-2	0	-3	-2	2	0	4	6									
N	0	-4	2	1	-4	0	2	-2	1	-3	-2	2								
P	1	-3	-1	-1	-5	-1	0	-2	-1	-3	-2	-1	6							
Q	0	-5	2	2	-5	-1	3	-2	1	-2	-1	1	0	4						
R	-2	-4	-1	-1	-4	-3	2	-2	3	-3	0	0	0	1	6					
S	1	0	0	0	-3	1	-1	-1	0	-3	-2	1	1	-1	0	2				
T	1	-2	0	0	-3	0	-1	0	0	-2	-1	0	0	-1	-1	1	3			
V	0	-2	-2	-2	-1	-1	-2	4	-2	2	2	-2	-1	-2	-2	-1	0	4		
W	-6	-8	-7	-7	0	-7	-3	-5	-3	-2	-4	-4	-6	-5	2	-2	-5	-6	17	
Y	-3	0	-4	-4	7	-5	0	-1	-4	-1	-2	-2	-5	-4	-4	-3	-3	-2	0	10

PAM 250

20x20 matrices

Point Accepted Mutations

A positive score means that a given *aa* substitution is favorable

A negative score means that a given *aa* substitution is unfavorable

	A	C	D	E	F	G	Н	Ι	K	L	M	N	P	Q	R	S	T	V	W	Y
A	4																			
C	0	9	-3																	
D	-2	-3	6																	
E	-1	-4	2	5																
F	-2	-2	-3	-3	6															
G	0	-3	-1	-2	-3	6														
Н	-2	-3	-1	0	-1	-2	8													
I	-1	-1	-3	-3	0	-4	-3	4												
K	-1	-3	-1	1	-3	-2	-1	-3	5											
L	-1	-1	-4	-3	0	-4	-3	2	-2	4										
M	-1	-1	-3	-2	0	-3	-2	1	-1	2	5									
N	-2	-3	1	0	-3	0	1	-3	0	-3	-2	6								
P	-1	-3	-1	-1	-4	-2	-2	-3	-1	-3	-2	-2	7							
Q	-1	-3	0	2	-3	-2	0	-3	1	-2	0	0	-1	5						
R	-1	-3	-2	0	-3	-2	0	-3	2	-2	-1	0	-2	1	5					
S	1	-1	0	0	-2	0	-1	-2	0	-2	-1	1	-1	0	-1	4				
T	0	-1	-1	-1	-2	-2	-2	-1	-1	-1	-1	0	-1	-1	-1	1	5			
V	0	-1	-3	-2	-1	-3	-3	3	-2	1	1	-3	-2	-2	-3	-2	0	4		
W	-3	-2	-4	-3	1	-2	-2	-3	-3	-2	-1	-4	-4	-2	-3	-3	-2	-3	11	
Y	-2	-2	-3	-2	3	-3	2	-1	-2	-1	-1	-2	-3	-1	-2	-2	-2	-1	2	7

BLOSUM 62

BLOcks **SU**bstitution **M**atrix (from the derived BLOCKS database)

A positive score means that a given *aa* substitution is favorable

A negative score means that a given *aa* substitution is unfavorable

PAM e BLOSUM matrices report the log₂ of:

 $\frac{f_{ij}}{f_i \times f_j}$

 \mathbf{f}_{ij} represents the frequency with whom two amino acids are found aligned

ex. f_{AT} is the number of times we observe an Ala aligned with a Thr, divided by the total number of pairs in an alignment

where

 \boldsymbol{f}_{i} and \boldsymbol{f}_{j} rapresent the frequencies with whom the two amino acids appear in an alignment

ex. f_A and f_T are the number of times we observe an Ala or a Thr, divided by the total number of amino acids in an alignment

Example of score calculation for the substitution of an Ala with a Thr

$$\begin{array}{c|c} f_{ij} & Seq1 & \texttt{CADGCFTL} \\ \hline f_i \ x \ f_j & Seq2 & \texttt{CTCGHILM} \\ \\ Seq3 & \texttt{TLCGHIAN} \end{array}$$

Tot aa =
$$3 \times 8 = 24$$

Tot aligned aa pairs = $3 \times 8 = 24$

$$\begin{aligned} f_{AT} &= 2 / 24 = 0.083 \\ f_{A} &= 2 / 24 = 0.083 \end{aligned} \qquad \begin{aligned} \frac{f_{AT}}{f_{A} x f_{T}} &= \frac{0.083}{0.010} = 8 \\ f_{T} &= 3 / 24 = 0.12 \\ f_{A} x f_{T} &= 0.010 \end{aligned} \qquad \begin{aligned} \frac{f_{AT}}{f_{A} x f_{T}} &= \frac{0.083}{0.010} = 8 \end{aligned}$$

Example of score calculation for the substitution of an Ala with a Thr

Tot aa = $3 \times 8 = 24$

Tot aligned aa pairs = $3 \times 8 = 24$

In a substitution matrix we would write 3 at the cross between Ala(A) and Thr(T)

$$\frac{f_{AT}}{f_A x f_T} = \frac{0.083}{0.010} = 8$$

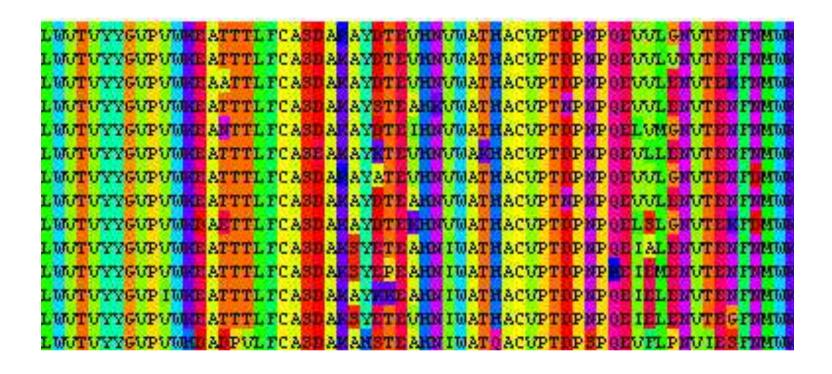
$$\ln_2 \left(\frac{f_{AT}}{f_A x f_T} \right) = 3$$

More rigorously...

$$s(ij) = int \left[k \cdot ln_2 \frac{f_{ij}}{f_i x f_j} \right]$$

We only take the integer

Site-specific matrices, based on empirical rules



PAM N: Percent/Point Accepted Mutations (where N is the number of accepted mutations every 100 aa)

BLOSUM N: BLOcks SUbstitution Matrix (where N is the maximum % of sequence identity between aligned homologs)

 PAM 1 can be used to generate matrices for higher evolutionary distances:

multiplying it again and again by itself.

PAM2 = PAM1 * PAM1

• PAM250:

etc etc

- 2,5 mutations per residue
- Equivalent to 20% remaining matches between two sequences, that is I'80% of amino acid positions are changed.
- It is the default matrix used in many analysis software.

- BLOSUM matrices have been developed to align scarcely correlated sequences. They have largely replaced the PAM ones.
- They are obtained from the derived **BLOCKS**databank containing alignments of highly
 correlated protein <u>regions</u>, which can be
 aligned without gaps.
- **BLOSUM62**: is obtained from alignments of proteine sharing a maximum of 62 % sequence identity. It is largely used. (Corrispondes approximately to a PAM110).

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More recently, matrices have been constructed using newer and larger data sets. The PET91 matrix, e.g., represents a new generation of Dayhoff-type matrices

	A	C	D	E	F	G	Н	I	K	L	M	N	P	Q	R	S	T	V	W	Y
A	2																			
C	-2	12																		
D	0	-5	4																	
Е	0	-5	3	4																
F	-4	-4	-6	-5	9															
G	1	-3	1	0	-5	5														
Н	-1	-3	1	1	-2	-2	6													
I	-1	-2	-2	-2	1	-3	-2	5												
K	-1	-5	0	0	-5	-2	0	-2	5											
L	-2	-6	-4	-3	2	-4	-2	2	-3	6										
M	-1	-5	-3	-2	0	-3	-2	2	0	4	6									
N	0	-4	2	1	-4	0	2	-2	1	-3	-2	2								
P	1	-3	-1	-1	-5	-1	0	-2	-1	-3	-2	-1	6							
Q	0	-5	2	2	-5	-1	3	-2	1	-2	-1	1	0	4						
R	-2	-4	-1	-1	-4	-3	2	-2	3	-3	0	0	0	1	6	_				
S	1	0	0	0	-3	1	-1	-1	0	-3	-2	1	1	-1	0	2	_			
T	1	-2	0	0	-3	0	-1	0	0	-2	-1	0	0	-1	-1	1	3			
V	0	-2	-2	-2	-1	-1	-2	4	-2	2	2	-2	-1	-2	-2	-1	0	4		
W	-6	-8	-7	-7	0	-7	-3	-5	-3	-2	-4	-4	-6	-5	2	-2	-5	-6	17	
Y	-3	0	-4	-4	7	-5	0	-1	-4	-1	-2	-2	-5	-4	-4	-3	-3	-2	0	10

PAM 250

20x20 matrices

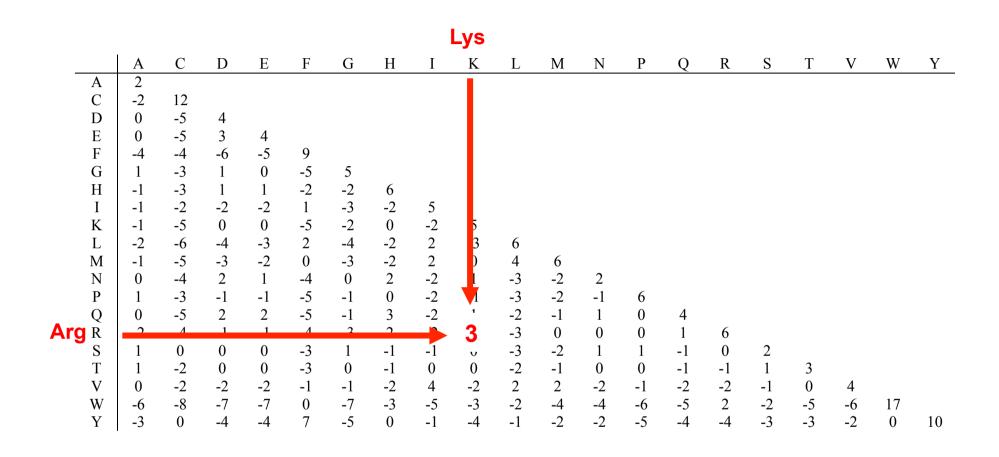
Point Accepted Mutations

	Ala																			
	Į,	С	D	Е	F	G	Н	I	K	L	M	N	P	Q	R	S	T	V	W	Y
A	2																	<u> </u>		
C		12																		
D	0	-5	4																	
Е	0	-5	3	4	0															
F G	-4 1	-4 2	-6 1	-5 0	9 -5	5														
Н	-1	-3 -3	1	0 1	-3 -2	5 -2	6													
I	-1	-2	-2	-2	1	-3	-2	5												
K	-1	-5	0	0	-5	-2	0	-2	5											
L	-2	-6	-4	-3	2	-4	-2	2	-3	6										
M	-1	-5	-3	-2	0	-3	-2	2	0	4	6									
N	0	-4	2	1	-4	0	2	-2	1	-3	-2	2								
P	1	-3	-1	-1	-5	-1	0	-2	-1	-3	-2	-1	6							
Q	0	-5	2	2	-5	-1	3	-2	1	-2	-1	1	0	4	(
R S	-2 1	-4 0	-1 0	-1 0	-4 -3	-3 1	2 -1	-2 -1	3	-3 -3	0 -2	0 1	0 1	1 -1	6 0	2				
T	1	-2	0	0	-3	0	-1 -1	0	0	-3 -2	-2 -1	0	0	-1 -1	-1	1	3			
V	0	-2	-2	-2	-1	-1	-2	4	-2	2	2	-2	-1	-2	-2	-1	0	4		
W	-6	-8	<u>-</u> 7	<u>-</u> 7	0	-7	-3	-5	-3	-2	<u>-4</u>	<u>-4</u>	-6	-5	2	-2	-5	-6	17	
Y	-3	0	-4	-4	7	-5	0	-1	-4	-1	-2	-2	-5	-4	-4	-3	-3	-2	0	10

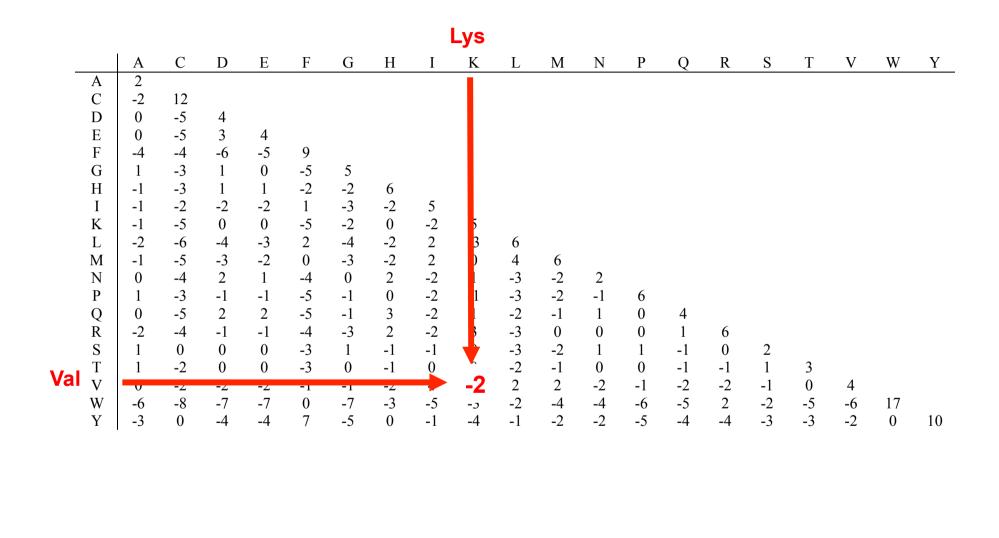
Ala residues are easily substituted by other aa

		Cys																		
		1																		
	A	C	D	Е	F	G	Н	I	K	L	M	N	P	Q	R	S	T	V	W	Y
A	2																			
C	-2	12																		
D	0	-5	4																	
Е	0	-5	3	4																
F	-4	-4	-6	-5	9															
G	1	-3	1	0	-5	5														
Н	-1	-3	1	1	-2	-2	6	_												
I	-1	-2	-2	-2	1_	-3	-2	5	_											
K	-1	-5	0	0	-5	-2	0	-2	5	_										
L	-2	-6	-4	-3	2	-4	-2	2	-3	6	_									
M	-1	-5	-3	-2	0	-3	-2	2	0	4	6	_								
N	0	-4	2	1	-4 -	0	2	-2	1	-3	-2	2								
P	1	-3	-1	-1	-5	-1	0	-2	-1	-3	-2	-1	6							
Q	0	-5	2	2	-5	-1	3	-2	1	-2	-1	1	0	4						
R	-2	-4	-1	-1	-4	-3	2	-2	3	-3	0	0	0	l	6	•				
S	1	0	0	0	-3	1	-1	-1	0	-3	-2	1	1	-1	0	2	2			
T	1	-2	0	0	-3	0	-1	0	0	-2	-1	0	0	-1	-1	l	3	4		
V	0	-2	-2	-2	-1	-1	-2	4	-2	2	2	-2	-1	-2	-2	-1	0	4	1.7	
W	-6	-8	-7	-7	0	-7	-3	-5 1	-3	-2	-4 2	-4 2	-6 -	-5 4	2	-2	-5 2	-6 2	17	1.0
Y	-3	0	- 4	- 4	7	-5	0	-1	- 4	-1	-2	-2	-5	-4	-4	-3	-3	-2	0	10

Cys residues are not easily substituted (they often give disulfide bonds)



Arg & Lys tend to substitute each other



Polar & apolar aa do not tend to substitute each other

Substitution matrices have been derived from alignments that <u>did not present insertions/</u> deletions (INDELs). Indels need therefore to be dealt with separately, on an empirical basis.

In aligning two sequences an Igorithm would tend to maximize the score (correspondence between identical or similar amino acids) by inserting a large number of gaps.

Is this the way which best reflects evolution?

We have indels (gaps) when a letter of a stretch of letters in one sequence is paired up with blanks spaces in another one

In nature INDEL events are often lethal (deleterious)

Therefore we need to penalize insertions and deletions. That means associating to them a negative score to be subtracted to the total score of the alignment.

In nature deletion of a series of contiguous nucleobases/amino acids is a more likely event than the independent deletion of the same number of nucleobases/amino acids in non contiguous positions

Let's distinguish the start of (introducing) a gap :

EGQTCA

AG-TCL

from the extension of (extending) a gap:

EGQQQTCA

AG---TCL

In nature deletion of a series of contiguous nucleobases/amino acids is a more likely event than the independent deletion of the same number of nucleobases/amino acids in non contiguous positions

Let's distinguish the start of a gap:

EGQTCA

AG-TCL

from the extension of a gap:

EGQQQTCA

AG---TCL

We penalize more a gap start than a gap extension

example:

-11 start

-1 extension

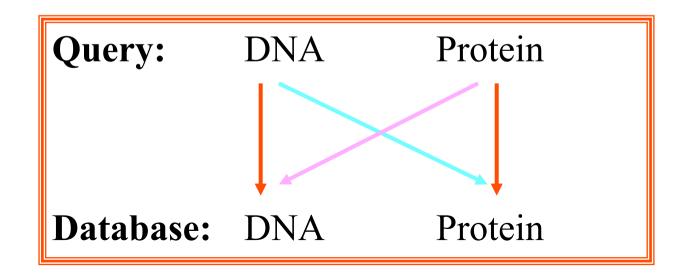
Sequence alignment – Part 1

For obtaining an optimal alignment (the one with maximum score, not necessarily the correct one, reflecting the evolutionary process), we need:

- A score for the substitution of amino acids/ nucleobases
- 2) Penalty for insertions/deletions (INDELs)
- 3) Algorithm to perform the alignment
- 4) Measure of the alignment significance

Homology search in databases

- Protein vs. proteins
- Gene (tranlastion to aa) vs. proteins
- Gene vs. genes
- Protein vs. translation to aa of nucleotide sequences (all frames)



When we compare protein sequences we search for the best correspondence for 20 different amino acids

When we compare nucleotide sequences we search for the best correspondence for only 4 nucleotides (nucleobases) When we compare protein sequences we search for the best correspondence for 20 different amino acids

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Probability of finding a good correspondence (high score alignment) by chance is higher for nucleotide sequences than for protein sequences

Furthermore, when we compare protein sequences we can take into account the similarity between amino acids

When we compare protein sequences we search for the best correspondence for 20 different amino acids

When we compare nucleotide sequences we search for the best correspondence for only 4 nucleotides (nucleobases)

Probability of finding a good correspondence (high score alignment) by chance is higher for nucleotide sequences than for protein sequences

Furthermore, when we compare protein sequences we can take into account the similarity between amino acids



When possible, comparing protein sequences has to be preferred!

How we can "fish" from databases potentially homologous sequences?



Exact algorithms (Smith-Waterman)

Exact, it provides the best alignment(s) for a pair of sequences.

Given 2 sequences: A of length n and B of length m, Smith-Waterman takes n*m computational steps.

If we search for homologs of the query sequence A (n=200 aa)

In a database made of 10⁶ sequences with m=200 aa

The number of computational steps is = $10^6 \times 200 \times 200 = -10^{10}$

 10^3 steps per sec = 10^7 secs = 120 days = 4 months!

There is a need for approximate (heuristic) algorithms

Exact algorithms (Smith-Waterman)

Exact, it provides the **best** alignment(s) for a pair of sequences.

Given 2 sequences: A of length n and B of length m, Smith-Waterman takes n*m computational steps.

How do we discard irrelevant alignments?

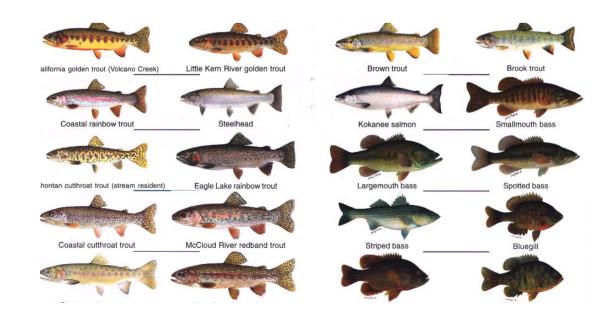


Heuristic algorithms (BLAST, FASTA) are needed to discard most of the irrelevant alignments.

Software such as **FASTA** and **BLAST**, starting from a query sequence:

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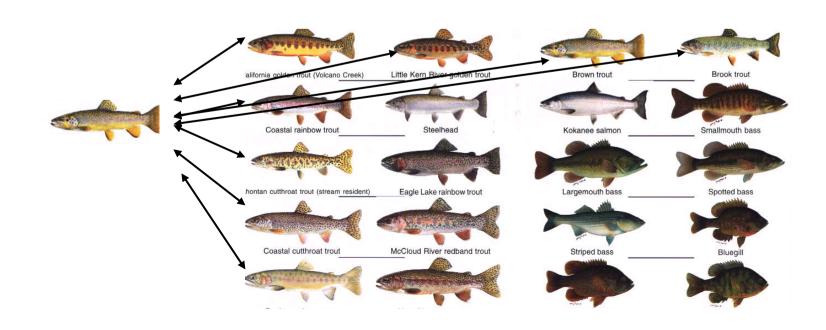
first "fish" from databases a subset of sequences which are potential homologs



Software such as **FASTA** and **BLAST**, starting from a query sequence:

first "fish" from databases a subset of sequences which are potential homologs

then perform the best alignment of each sequence in the subset with the query sequence



FASTA: example

query
ACDDEFGSATRMASTRK

Data bank (DB)

Seq 1: LKDCDDAFSGSTLTLMRASRK

Seq 2: ACKRAEFSGSVTRMLSTRK

Seq 3: ACDDEFGLLLTRYTMASTRK

Step 1 = Division of the sequence in 2-letter words (k-tuples).

Possible words:

AC, CD, DD, DE, EF, FG, GS, SA, AT, TR, RM, MA, AS, ST, RK

Note. A typical value of k for DNA is 6

Step 2 = Table of word frequencies

Query: ACDDEFGSATRMASTRK DB: Seq 1 LKDCDDAFSGSTLTLMRASRK

Seq 2 ACKRAEFSGSVTRMLSTRK

Seq 3 ACDDEFGLLLTRYTMASTRK

Word	Query	Seq 1	Seq 2	Seq 3	Off1	Off2	Off3
AC	1	-	1	1	-	0	0
CD	2	4	_	2	2	-	0
DD	3	5	-	3	2	-	0
DE	4	_	_	4	-	-	0
EF	5	_	6	5	-	1	0
FG	6	_	_	6	-	-	0
GS	7	10	9	-	3	2	-
SA	8	_	-	-	-	-	-
AT	9	_	-	_	-	-	-
TR	10	_	12, 17	11	-	2 , 7	1
RM	11	_	13	_	-	2	-
MA	12	_	_	15	-	-	3
AS	13	18	-	16	5	-	3
ST	14	11	16	17	-3	2	3
RK	16	20	18	19	4	2	3

Step 3 = Similarity score calculation *Init1* (based on the Table at the step 2)

Query: ACDDEFGSATRMASTRK DB: Seq 1 LKDCDDAFSGSTLTLMRASRK

Seq 2 ACKRAEFSGSVTRMLSTRK

Seq 3 ACDDEFGLLLTRYTMASTRK

Query		/	4	С	D	D	Ε	F	G	S	Α	Т	R	M	Α	S	Т	R	K	
Pos		•	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	
									X	X		X	X			X	X	X	X	
Seq2	Α (C F	<	R	Α	Е	F	S	G	S	V	Т	R	M	L	S	Т	R	K	Т
Pos	1 2	2 (3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
Off									2	2		2	2			2	2	2	2	

Init1(seq2) based on this approximate alignment

Table of word frequencies

Word AC	Query 1	Seq 1	Seq 2 1	Seq 3	Off1	Off2 0	Off3
CD	2	4	_	2	2	_	0
DD	3	5	-	3	2	-	0
DE	4	_	_	4	_	-	0
EF	5	_	6	5	-	1	0
FG	6	_	_	6	_	-	0
GS	7	10	9	_	3	2	-
SA	8	_	_	_	_	-	-
AT	9	_	_	_	_	-	-
TR	10	_	12, 17	11	_	2 , 7	1
RM	11	_	13	_	-	2	-
MA	12	_	_	15	_	-	3
AS	13	18	-	16	5	-	3
ST	14	11	16	17	-3	2	3
RK	16	20	18	19	4	2	3

Step 4 = Similarity score calculation *InitN* (based on the alignment at step 3 and on the Table at step 2)

Query: ACDDEFGSATRMASTRK DB: Seq 1 LKDCDDAFSGSTLTLMRASRK

Seq 2 ACKRAEFSGSVTRMLSTRK

Seq 3 ACDDEFGLLLTRYTMASTRK

Query	A C -	D	D	Е	F	-	G	S	Α	Т	R	M	Α	S	Т	R	K	
Pos	1 2	3	4	5	6		7	8	9	10	11	12	13	14	15	16	17	
	ХХ			X	X		X	X		X	X			X	X	X	X	
Seq2	ACK	R	A	Ε	A	S	G	S	V	Т	R	M	L	S	Т	R	K	Т
Pos	1 2 3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19	20
Off	0									4	2			2	2	2	2	

InitN(seq2) = Init1(seq2) + score(novel matches) - K(gap)

Step 5 = Final alignment of the sequences with the best *InitN* score with the query sequence and calculation of the final score opt (score for the novel, complete alignment)

NOTE. The <u>choice of the sequences subset</u> in the DB with whom the optimal alignment is finally performed is based on the approximate scores Init1 & InitN

FASTA

- 1. Divides the query sequence in 2-letter words (k-tuples).
- 2. Finds these words in the database sequences and calculates the offset
- 3. Calculates the similarity of the ten regions with most identical words for each sequence in the DB (init1)
- 4. Calculates the similarity of the ten regions with most identical words including penalization for insertions & deletions
- Accurately aligns the N sequences with best initN score → obtaining opt

How good an alignment is?

How good an alignment is?

How better than a random alignment it is?

(Unrelated) sequences which give a random alignment:

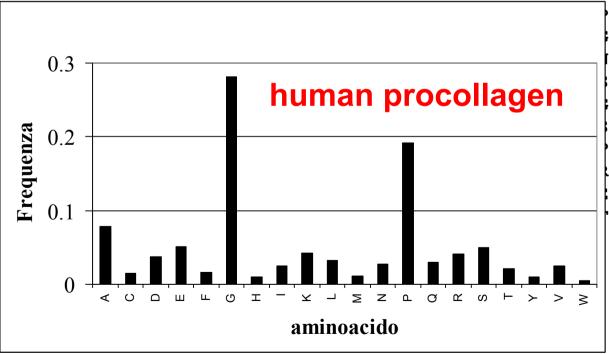
- Non-homologous sequences
- Shuffled sequences
- Randomly generated sequences
- Low complexity sequences

Low complexity

MMSFVOKGSW LLLALLHPTI ILAQOEAVEG GCSHLGOSYA DRDVWKPEPC OICVCDSGSV LCDDIICDDO ELDCPNPEIP FGECCAVCPQ PPTAPTRPPN GQGPQGPKGD PGPPGIPGRN GDPGIPGQPG SPGSPGPPGI CESCPTGPON YSPOYDSYDV KSGVAVGGLA GYPGPAGPPG PPGPPGTSGH PGSPGSPGYO GPPGEPGOAG PSGPPGPPGA IGPSGPAGKD GESGRPGRPG ERGLPGPPGI KGPAGIPGFP GMKGHRGFDG RNGEKGETGA PGLKGENGLP GENGAPGPMG PRGAPGERGR PGLPGAAGAR GNDGARGSDG OPGPPGPPGT AGFPGSPGAK GEVGPAGSPG SNGAPGORGE PGPOGHAGAO GPPGPPGING SPGGKGEMGP AGIPGAPGLM GARGPPGPAG ANGAPGLRGG AGEPGKNGAK GEPGPRGERG EAGIPGVPGA KGEDGKDGSP GEPGANGLPG AAGERGAPGF RGPAGPNGIP GEKGPAGERG APGPAGPRGA AGEPGRDGVP GGPGMRGMPG SPGGPGSDGK PGPPGSOGES GRPGPPGPSG PRGOPGVMGF PGPKGNDGAP GKNGERGGPG GPGPOGPPGK NGETGPOGPP GPTGPGGDKG DTGPPGPOGL OGLPGTGGPP GENGKPGEPG PKGDAGAPGA PGGKGDAGAP GERGPPGLAG APGLRGGAGP PGPEGGKGAA GPPGPPGAAG GPAGQPGDKG EGGAPGLPGI VAGPPGGSGP AGPPGPOGVK 0.3 TGAPGSPGVS GPKGDAGOPG GKPGANGLSG ERGPPGPQGL PPGPVGPAGK SGDRGESGPA 0.2 PGPAGQQGAI GSPGPAGPRG GAPGPCCGGV GAAAIAGIGG

NCRDLKFCHP ELKSGEYWVD

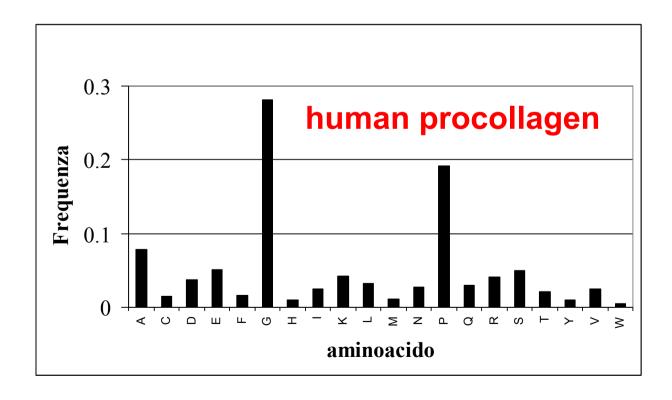
SMDGGFQFSY GNPELPEDVL KAEGNSKFTY TVLEDGCTKH



Low complexity

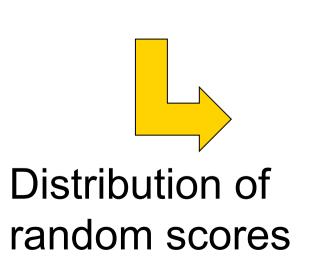
Low complexity regions in protein sequences have a highly biased amino acid composition, often repeats of proline, alanine, serine, glycine, leucine, and glutamic acid

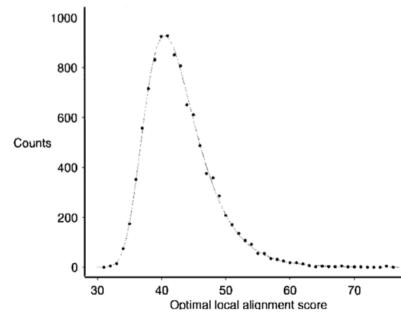
Especially abundant in eukaryotic proteins



Are they homologous sequences? Evaluating the significance of the alignment

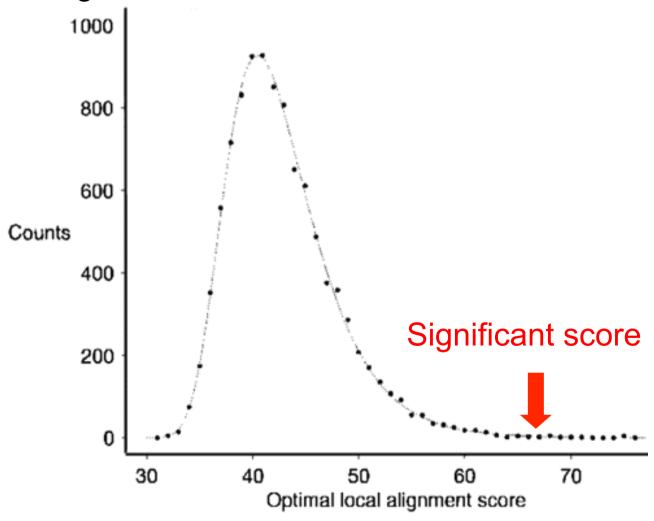
- a) Generating a large number of random sequences with the same composition of the query seq ("shuffled" sequences)
- b) Ripeting the similarity search on random of the DBs using as a query each of the random sequence
- c) Calculating corresponding opt scores, their average value M_{random} and their standard devation σ_{random}





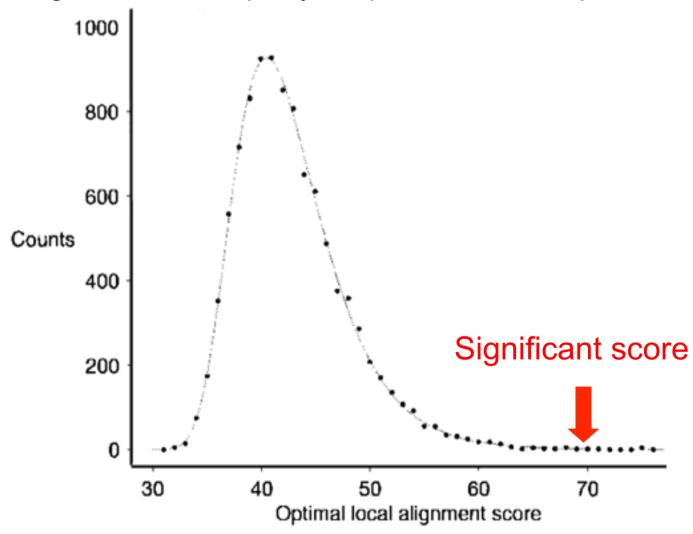
Evaluating the significance of the alignment

Two sequences can be considered homologous if the optimal score (opt) for their alignment falls off the random scores distribution



Evaluating the significance of the alignment

4. Calculating the *Z-score* and the expectation value (*E-value*) for the alignment of the query sequence with its putative homologs

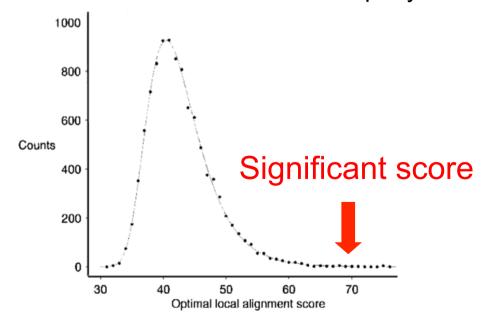


Evaluating the significance of the alignment

4. Calculating the **Z-score** and the expectation value (*E-value*) for the alignment of the query sequence with its putative homologs

Z-score = number of standard deviations which separate the query score (opt) from the average of the random scores

Z-score (S) =
$$(opt_{query} - M_{random}) / \sigma_{random}$$



Z-score » 4

→ opt_{query} off the random distribution

average

standard deviation

$$M_{\text{random}} = \frac{\sum_{i} (opt_i)}{n}$$

$$\sigma_{\text{random}} = \sqrt{\frac{\sum\limits_{i} (opt_i - M_{\text{random}})}{n-1}}$$

Evaluating the significance of the alignment

E-value = expectation value: number of alignments with a score ≥ S (o opt) that would be expected by chance by searching a complete database of size *n* (length of all sequences)

> Indicates how probable is finding a score S by chance

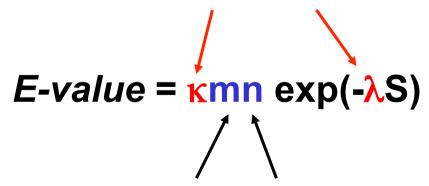
E-value = κmn exp(- λ S)

Evaluating the significance of the alignment

E-value = expectation value: number of alignments with a score ≥ S (o opt) that would be expected by chance by searching a complete database of size n (length of all sequences)

The lower, the better!

Statistical parameters, dipending on the matrix and the DB



size of the query (m) size of the database (n)

Evaluating the significance of the alignment

E-value = expectation value: number of alignments with a score ≥ S (o opt) that would be expected by chance by searching a complete database

> The typical threshold for a good E-value from a FASTA/BLAST search is E=10⁻⁵ or lower

E-value = κmn exp(- λ S)

The **probability** of having by chance an alignment with a score \geq S is given by:

$$P = 1 - e^{(-\kappa mn \exp(-\lambda S))}$$

Evaluating the significance of the alignment

It is possible to normalize the score:

S' =
$$(\lambda S - \ln \kappa) / \ln 2$$

Bit-score
$$\Box$$
E-value = mn 2-S'

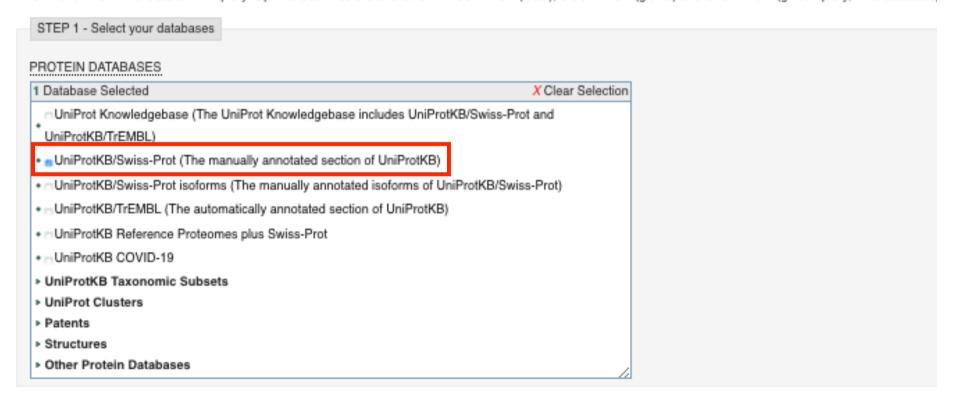
The bit-score is independent from query sequence length and database size

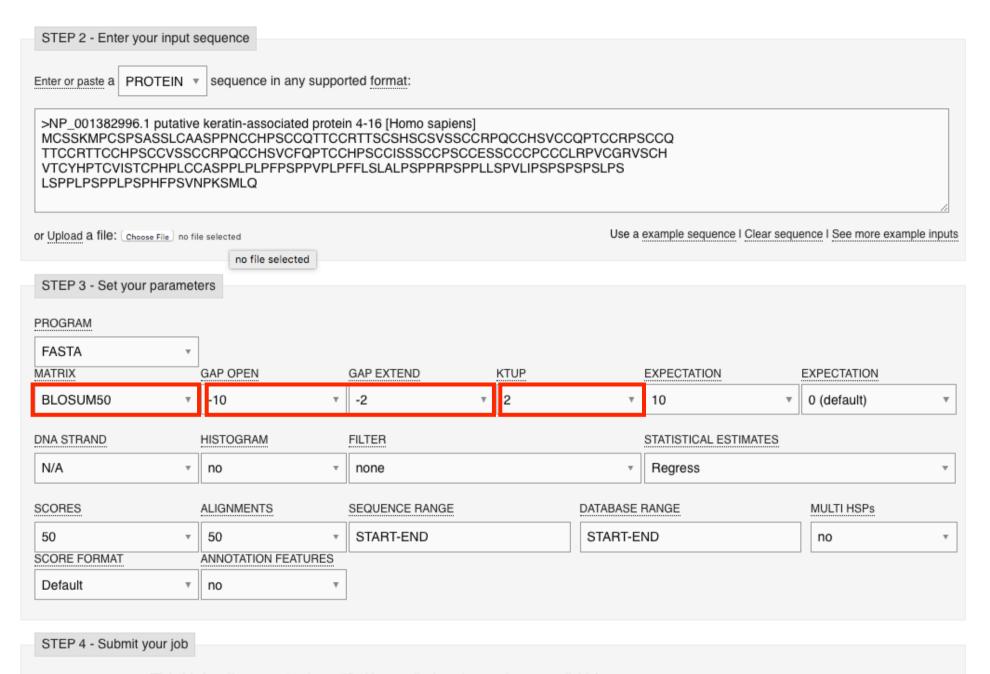
Thus, it is possible to compare directly the obtained bit-scores from searches in different databases and with different matrices

Tools > Sequence Similarity Searching > FASTA

Protein Similarity Search

This tool provides sequence similarity searching against protein databases using the FASTA suite of programs. FASTA provides a heuristic search with a protein query. FASTX and FASTY translate a DNA query. Optimal searches are available with SSEARCH (local), GGSEARCH (global) and GLSEARCH (global query, local database)





 $\ \square$ Be notified by email (Tick this box if you want to be notified by email when the results are available)

Submit

Tools > Sequence Similarity Searching > FASTA

Results for job fasta-I20220325-093358-0760-32105718-p2m

Summary Table Tool Output Visual Output Functional Predictions Submission Details

Select All Invert Clear	
Apply to selection:	
Annotations:	
Show Hide	
Alignments:	
Show	
Entries:	
Download in	
fasta	v
format	
Tools:	
Launch	
Clustal Omega	₩

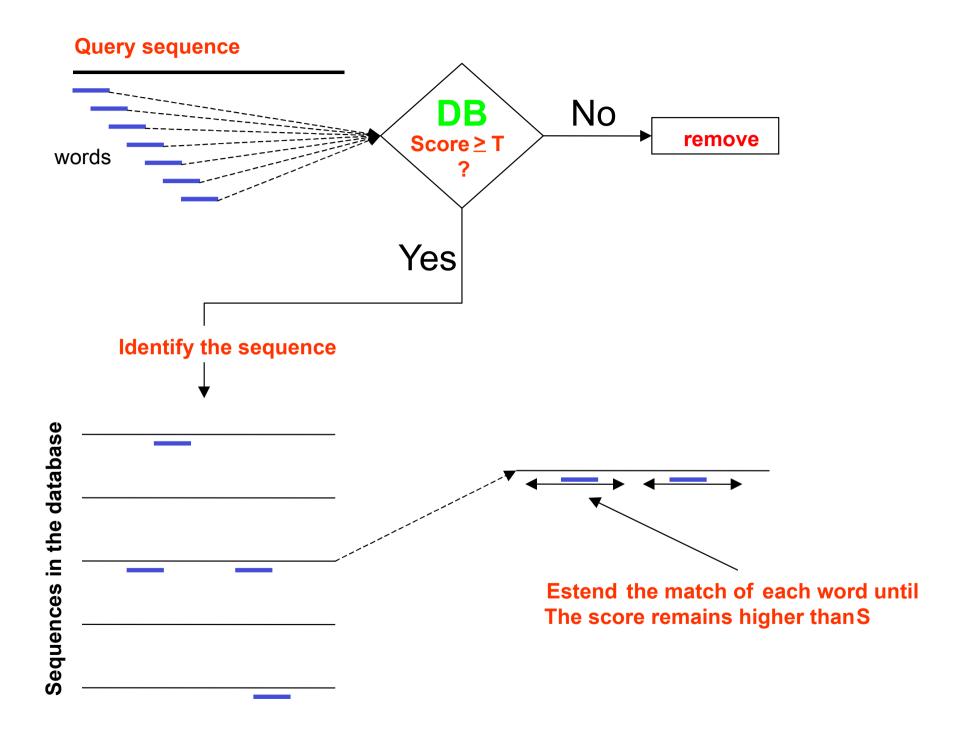
Align	DB:ID \$	Source	Length ≑	Score (Bits) ‡	Identities \$	Positives \$	E() \$
⊡ 1	SP:G5E9R7	Putative keratin-associated protein 4-16 OS=Homo sapiens OX=9606 GN=KRTAP4-16 PE=5 SV=1 Cross-references and related information in: ▶ Gene expression ▶ Nucleotide sequences ▶ Genomes & metagenomes ▶ Literature ▶ Samples & ontologies ▶ Protein families ▶ Protein expression data ▶ Protein sequences	235	219.0	100.0	100.0	3.8E-56
⊠ 2	SP:Q9BQ66	Keratin-associated protein 4-12 OS=Homo sapiens OX=9606 GN=KRTAP4-12 PE=1 SV=1 Cross-references and related information in: ▶ Gene expression ▶ Bioactive molecules ▶ Nucleotide sequences ▶ Genomes & metagenomes ▶ Literature ▶ Samples & ontologies ▶ Molecular interactions ▶ Protein families ▶ Protein expression data ▶ Protein sequences ▶ Diseases	201	133.6	81.4	89.4	1.6E-30
₽ 3	SP:Q9BYR2	Keratin-associated protein 4-5 OS=Homo sapiens OX=9606 GN=KRTAP4-5 PE=1 SV=4 Cross-references and related information in: ▶ Gene expression ▶ Bioactive molecules ▶ Nucleotide sequences ▶ Genomes & metagenomes ▶ Literature ▶ Samples & ontologies ▶ Molecular interactions ▶ Protein families ▶ Protein expression data ▶ Reactions & pathways ▶ Protein sequences ▶ Diseases	181	120.4	68.9	82.9	1.4E-26
2 4	SP:Q9BYQ8	Keratin-associated protein 4-9 OS=Homo sapiens OX=9606 GN=KRTAP4-9 PE=2 SV=2 Cross-references and related information in: ▶ Gene expression ▶ Bioactive molecules ▶ Nucleotide sequences ▶ Genomes & metagenomes ▶ Literature ▶ Samples & ontologies ▶ Protein families ▶ Protein expression data ▶ Reactions & pathways ▶ Protein sequences ▶ Diseases	210	119.7	69.9	83.1	2.5E-26
~ 5	SP:Q9BYQ5	Keratin-associated protein 4-6 OS=Homo sapiens OX=9606 GN=KRTAP4-6 PE=2 SV=4 Cross-references and related information in: ▶ Gene expression ▶ Bioactive molecules ▶ Nucleotide sequences ▶ Genomes & metagenomes ▶ Literature ▶ Samples & ontologies ▶ Protein families ▶ Protein expression data ▶ Reactions & pathways ▶ Protein sequences ▶ Diseases	205	117.4	64.6	71.2	1.2E-25
⊘ 6	SP:Q9BYQ6	Keratin-associated protein 4-11 OS=Homo sapiens OX=9606 GN=KRTAP4-11 PE=1 SV=2 Cross-references and related information in: ▶ Gene expression ▶ Bioactive molecules ▶ Nucleotide sequences	195	117.3	62.6	72.1	1.3E-25

```
>>SP:09B066 KR412 HUMAN Keratin-associated protein 4-12
OS=Homo sapiens OX=9606 GN=KRTAP4-12 PE=1 SV=1 (201 aa)
initn: 1297 init1: 1126 opt: 1142 Z-score: 683.6 bits: 133.6 E(566996): 1.6e-30
Smith-Waterman score: 1142; 81.4% identity (89.4% similar) in 161 aa overlap (2-161:40-198)
                                         10
                                                    20
NP 001
                                  MCSSKMPCSPSA-SSLCAASPPNCCHPSCCO
                                   : . : :. .:.:
SP:09B CSDOGCGLENCCRPSCCOTTCCRTTCCRPSCCVSSCCRPOCCOSVCCO--PTCCRPSCCO
               20
                        30
                                           50
     10
                                  40
                                                       60
                       50
                                 60
              40
                                          70
                                                    80
NP 001 TTCCRTTSCSHSCSVSSCCRPQCCHSVCCQPTCCRPSCCQTTCCRTTCCHPSCCVSSCCR
       SP:09B TTCCRTTCCRPSCCVSSCCRPOCCOSVCCOPTCCRPSCCOTTCCRTTCCRPSCCVSSCCR
       70
                 80
                          90
                                   100
                                            110
                                                      120
                                120
             100
                      110
                                         130
                                                   140
                                                            150
NP 001 PQCCHSVCFQPTCCHPSCCISSSCCPSCCESSCCCPCCCLRPVCGRVSCHVTCYHPTCVI
SP:09B POCCOSVCCOPTCCRPSCCISSSCCPSCCESSCCRPCCCLRPVCGRVSCHTTCYRPTCVI
                                            170
                140
                         150
      130
                                   160
                                                      180
                      170
             160
                                180
                                         190
                                                   200
                                                             210
NP 001 STCPHPLCCASPPLPLPFPSPPVPLPFFLSLALPSPPRPSPPLLSPVLIPSPSPSPSLPS
SP:09B STCPRPLCCASSCC
                200
      190
```

BLAST

(Basic Local Alignment Search Tool)

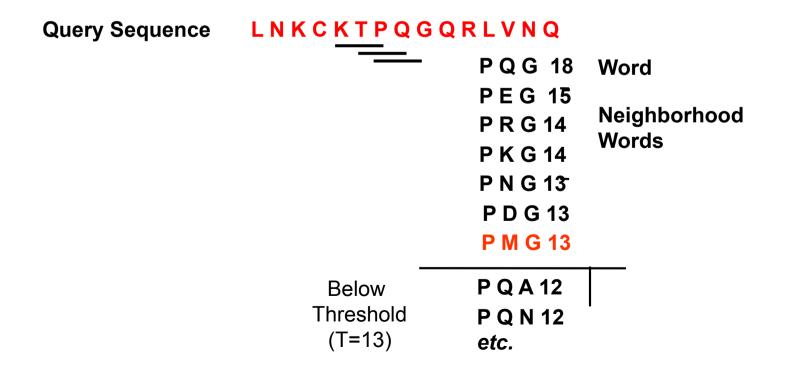
- 1. Divides the query sequence in words (*default*, 3 aa)
- 2. Compares each word with regions of same size in the DB sequences and computes the *score*
- 3. If the score is ≥ a threshold value T below which the similarity is considered too low, extends the aligned region searching for high similarity regions (score above a second threshold value S), stopping when the score cannot be improved anymore



BLAST Algorithm, Step 1

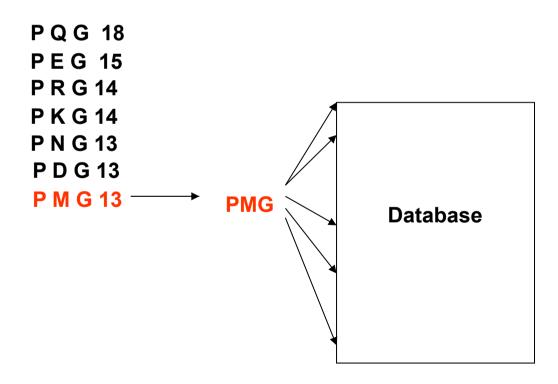
 Given a word of length w (usually 3 for proteins) & a scoring/substitution matrix (es. BLOSUM62):

Create a list of all the words (*w*-letters) that give a **score** >*T* sewhen compared with the query words of length *w*-



BLAST Algorithm, Step 2

 Identifies all the positions in the database where there is a word sufficiently similar (hit list).



BLAST Algorithm, Step 3

• The software attempts to **extend** the alignment in both directions adding pairs of residues. Residues are added until the score cannot be improved anymore. It considers only alignments with a score above the threshold value (S).

```
Query: 325 SLAALLNKCKTPQGQRLVNQWIKQPLMDKNRIEERLNLVEA 365
+LA++L+ TP G R++ +W+ P+ D + ER + A
Sbjct: 290 TLASVLDCTVTPMGSRMLKRWLHMPVRDTRVLLERQQTIGA 330
```

High-scoring Segment Pair (HSP)

High Scoring Segment Pairs

BLAST & **FASTA** differ in the way they "fish" putative homologs from the DB (similarity/identity).

BLAST & **FASTA** differ in the way they "fish" putative homologs from the DB (similarity/identity)

Furthermore a fundamental difference between **BLAST** & **FASTA** is in the way they compute the distribution of random scores:

FASTA computes it each time a novel query is submitted for the search in a given DB

BLAST uses distributions precomputed on each DB for ensembles of random sequences of standard composition

BLAST e **FASTA** differ in the way they "fish" putative homologs from the DB (similarity/identity)

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FASTA computes it each time a novel query is submitted for the search in a given DB

BLAST uses <u>distributions precomputed</u> on each DB for ensembles of random sequences of <u>standard composition</u>



For this reason BLAST "masks" the regions of query sequence at low complexity



Specialized searches

SmartBLAST

Find proteins highly similar to your query

IgBLAST

Search immunoglobulins and T cell receptor sequences

MOLE-BLAST

Primer-BLAST

Design primers specific to your PCR template

VecScreen

Search sequences for vector contamination

Global Align

Compare two sequences across their entire span (Needleman-Wunsch)

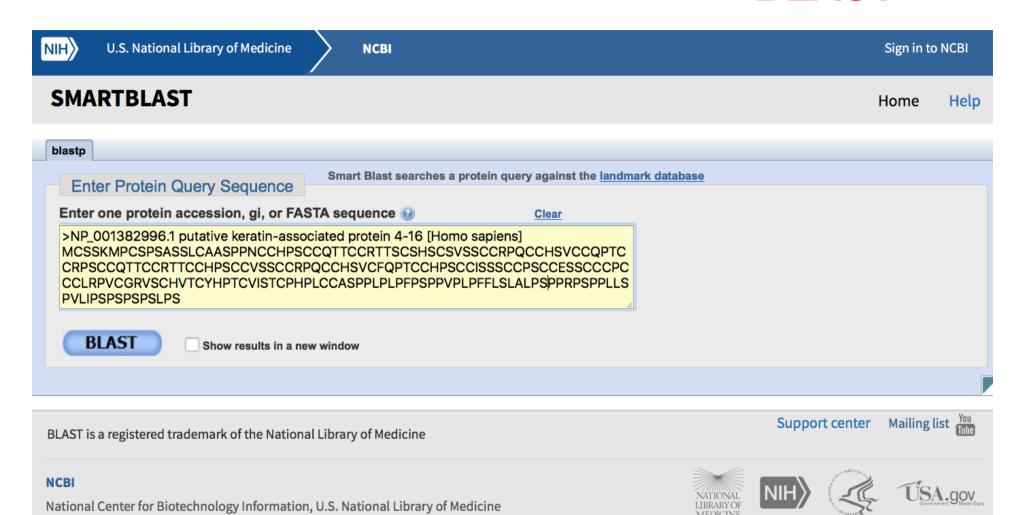
CDART

Find sequences with similar conserved domain architecture **CD-search**

Find conserved domains in your sequence

Multiple Alignment

Align sequences using domain and protein constraints



8600 Rockville Pike, Bethesda MD, 20894 USA

Best hits

Additional BLAST Hits

Select: All None Selected:0

Alignments GenPept											
	Description	Max score	Total score	Query	E value	Ident	Accession				
	putative keratin-associated protein 4-16 [Homo sapiens]	350	350	100%	6e-120	100.00%	NP_001382996.1				
	KRTAP4-16 isoform 1 [Pan troglodytes]	211	211	93%	2e-65	84.62%	PNI34085.1				
	keratin-associated protein 4-8 isoform X2 [Pongo abelii]	167	167	67%	1e-48	76.03%	XP_024091030.1				
	keratin-associated protein 4-9-like isoform X2 [Nomascus leucogenys]	164	164	67%	7e-47	75.50%	XP_012353014.1				
	keratin-associated protein 4-8 isoform X9 [Pan paniscus]	164	164	66%	1e-46	77.14%	XP_034799243.1				
	keratin-associated protein 4-11 isoform X9 [Pan troglodytes]	159	159	66%	2e-45	77.14%	XP_024206089.1				
	keratin-associated protein 4-9-like isoform X4 [Nomascus leucogenys]	157	157	57%	5e-45	82.40%	XP_004091436.1				
	keratin-associated protein 4-9-like isoform X1 [Nomascus leucogenys]	159	159	67%	5e-45	73.08%	XP_004091435.1				
	keratin-associated protein 4-11 isoform X6 [Pan paniscus]	157	157	66%	8e-44	69.68%	XP_034799240.1				
	keratin-associated protein 4-9-like isoform X3 [Nomascus leucogenys]	155	155	67%	1e-43	74.17%	XP_003278340.2				
	keratin-associated protein 4-9 isoform X3 [Pan paniscus]	155	155	66%	4e-43	68.75%	XP_034799237.1				
	keratin-associated protein 4-11 isoform X5 [Pan troglodytes]	153	153	66%	1e-42	69.68%	XP_003953071.3				
	keratin-associated protein 4-11 isoform X10 [Pan paniscus]	152	152	57%	3e-42	79.20%	XP_034799244.1				
	keratin-associated protein 4-12 isoform X4 [Cavia porcellus]	150	219	67%	6e-42	70.21%	XP_023419679.1				
1											

▼ Next ▲ Previous ▲ Descriptions GenPept

KRTAP4-16 isoform 1, partial [Pan troglodytes]

Sbjct 181 TLPLPFFLSLALPSPPHTSPPLLSTVLI 208

Sequence ID: PNI34085.1 Length: 228 Number of Matches: 1

Range 1: 1 to 208 GenPept	ept	GenP	208	to	1	1:	le	lang	R
---------------------------	-----	------	-----	----	---	----	----	------	---

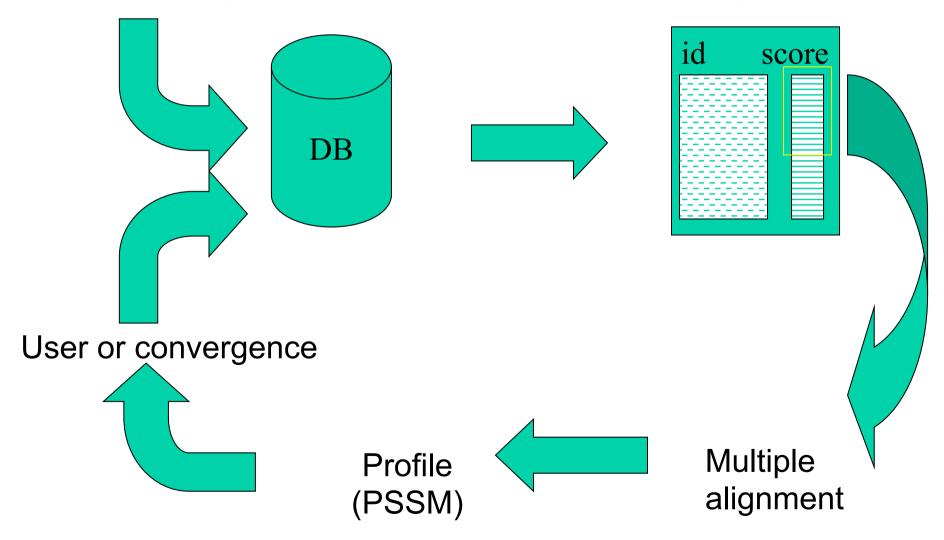
Range	1: 1 to	▼ Next Match	A Previous Mat	ch			
Score		Expect	Method	Identities	Positives	Gaps	Frame
211 b	its(538	3) 2e-65()	Compositional matrix adjust	st. 176/208(85%)	180/208(86%)	12/208(5%)	
Query	4		SSLCAASPPNCCHPSCCQTTCCRTT: SS+CAASPPNCCHPS COTTCCRTT:				
Sbjct	1		SSVCAASPPNCCHPSSCQTTCCRTT				
Query	64	-	CCRTTCCHPSCCVSSCCRPQCCHSV				
Sbjct	61	-	CCRMTCCHPSCCVSSCCRPQCCHSV	-			
Query	119		CLRPVCGRVSCHVTCYHPTCVIST		PLPLPFPSP 171 PLP PSP		
Sbjct	121	CESSCCCPFC	CLRPVCGRVSCHITCYHPTCVIST	CPRPLCCASPPLPLLSPS	SPPLPLPSP 180		
Query	172	PVPLPFFLSI +PLPFFLSI	ALPSPPRPSPPLLSPVLI 199 ALPSPP SPPLLS VLI				

Related Information

NewGenome Data Viewer aligned genomic context

PSI-blast

(Position-Specific Iterative blast)



PSI-blast

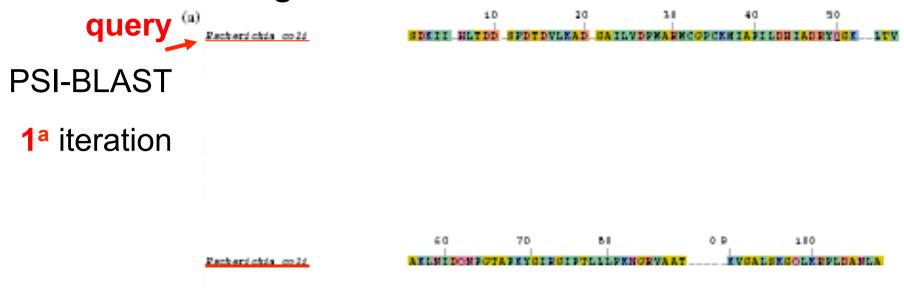
(Position-Specific Iterative blast)

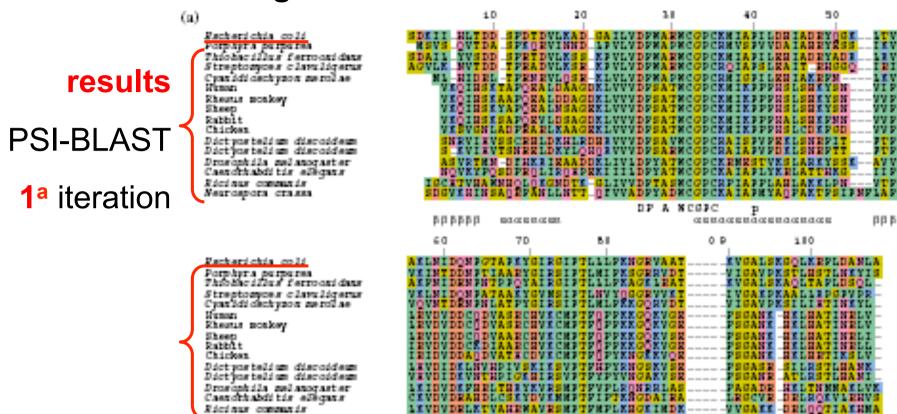
PSI-BLAST is a variation of BLAST that uses features of a particular protein family to identify related sequences in a protein database

In PSI-BLAST a profile or position-specific scoring matrix (PSSM) of a set of sequences is constructed from a multiple alignment of the highest-score hits found by the initial BLAST search

In the PSSM a high score is assigned to a highly conserved residue at a certain position while a negative score is assigned to other residues at that position

The profile generated is used to replace the substitution matrix in a subsequent BLAST search

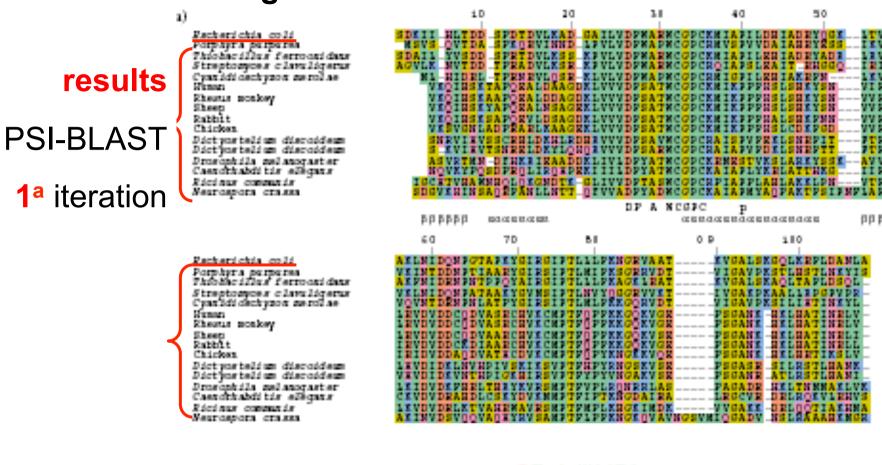




AKINYDSYQQVAQXYRYSANFTPLPPKNGKQYAVNG

SVMIQUADV NELEKARDENCE

Neurospora crassa





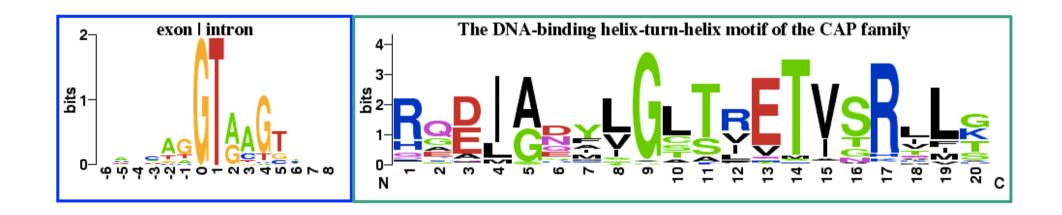
PSI-BLAST





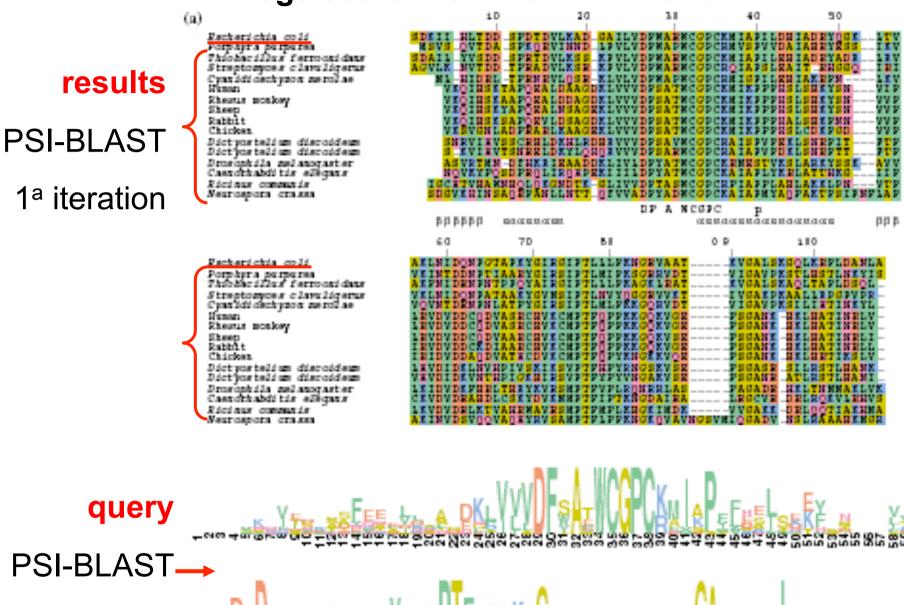
Sequence logos

Profiles of multiple sequence alignments can be represented graphically in the form of sequence logos, easily showing the residue preference or conservation at particular positions, which point to a functional role

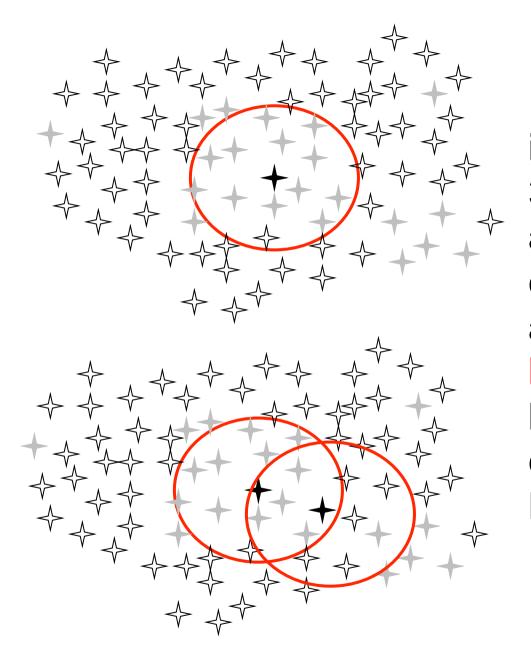


Examples from Web Logo

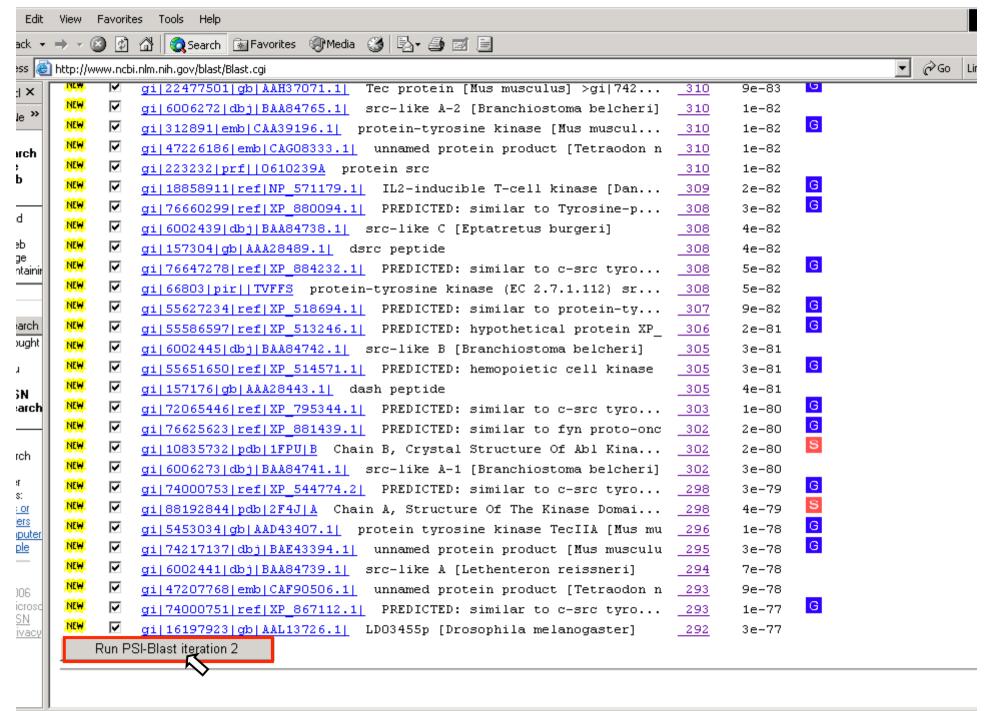
https://weblogo.berkeley.edu/logo.cgi

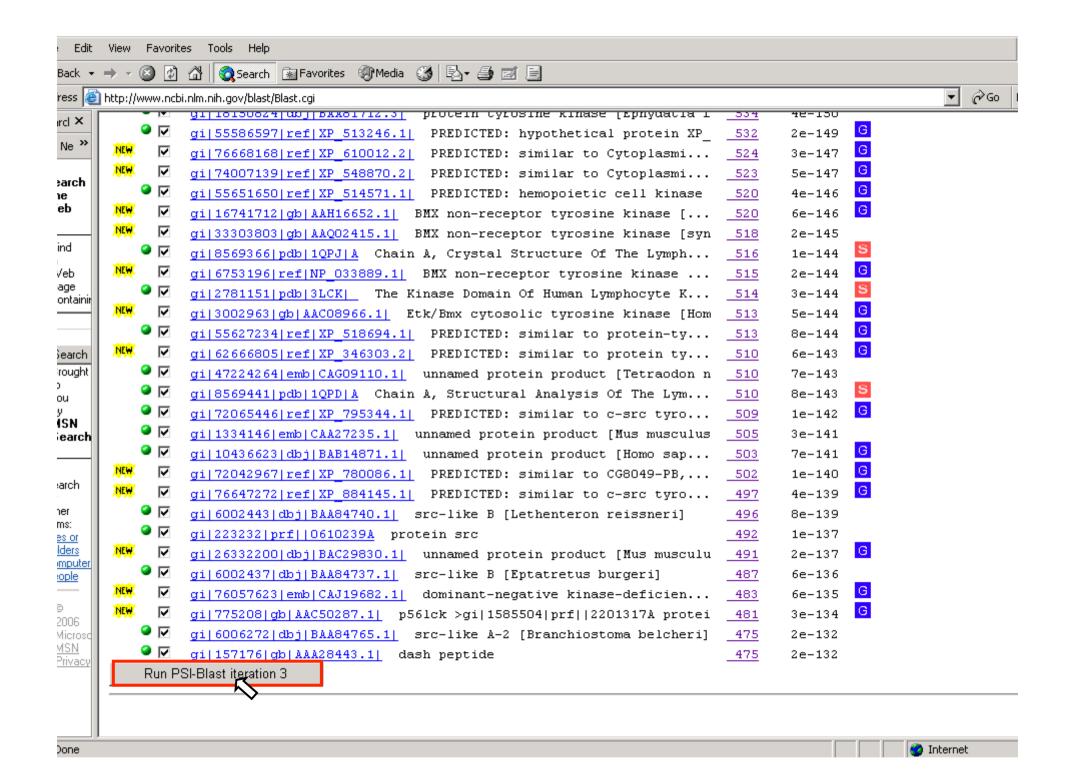


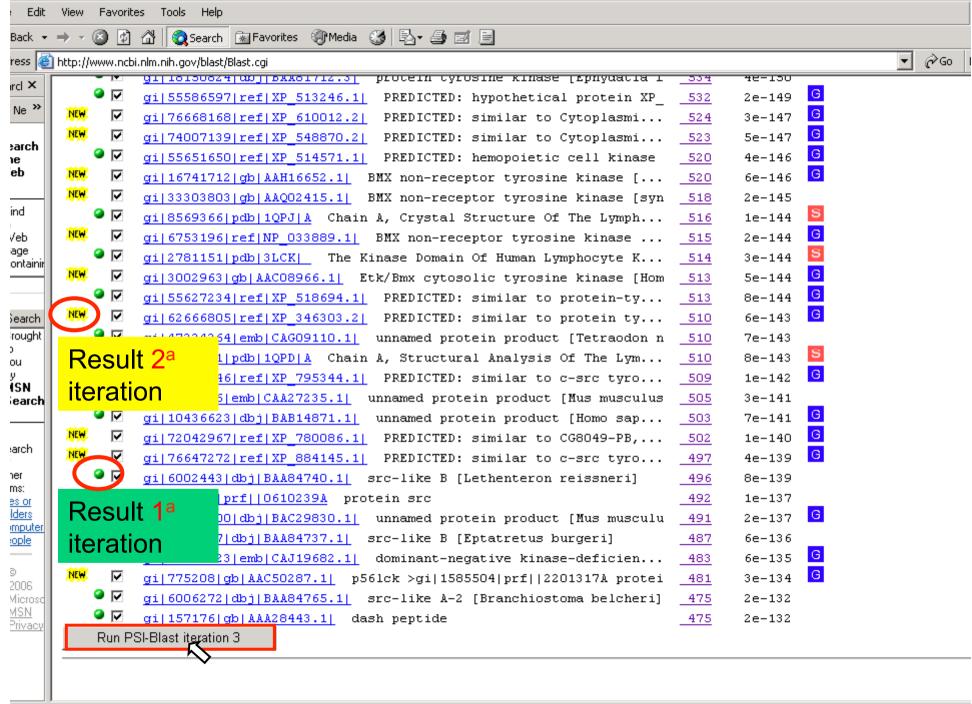
2^a iteration



For sequence identities below 30 % PSI-BLAST allows to correctly identify a three-fold higher number of homologs as as compared to **BLAST**







Are homologs found for all the protein sequences?

Unique sequences, i.e. sequences with no significant match in homologs searches (BLAST hit with E-value >10⁻³ or > 10⁻⁵ for alignments of < 80 residues) are referred to as orphan ORFs or ORFans and, in particular, singleton ORFans

The percentage of singleton ORFans in each newly sequenced genome can be as high as 60%

In addition to these unique ORFans, a large fraction of ORFs in each genome has homologs only in the same genome or in closely related genomes. These ORFs are referred to as paralogous and orthologous ORFans, respectively

Lessons 3 & 4. Contents

- 1. Introduction to proteins. Different sequences correspond to different 3D structures. Specific structures determine specific functions.
- 2. Sequence alignment. We search for those that best reflect the evolutionary path. Based on sequence similarity we can infer homology (an evolutionary relationship)
- 3. Substitutions & gaps. Substitution matrices allow to assign a score for the correspondence of different amino acids. It is necessary to penalize insertions and deletions (INDELs).
- 4. Homology search in databases. BLAST & FASTA identify a subset of sequences from the databanks, align them to the query and compare the obtained score to a distribution of random scores.