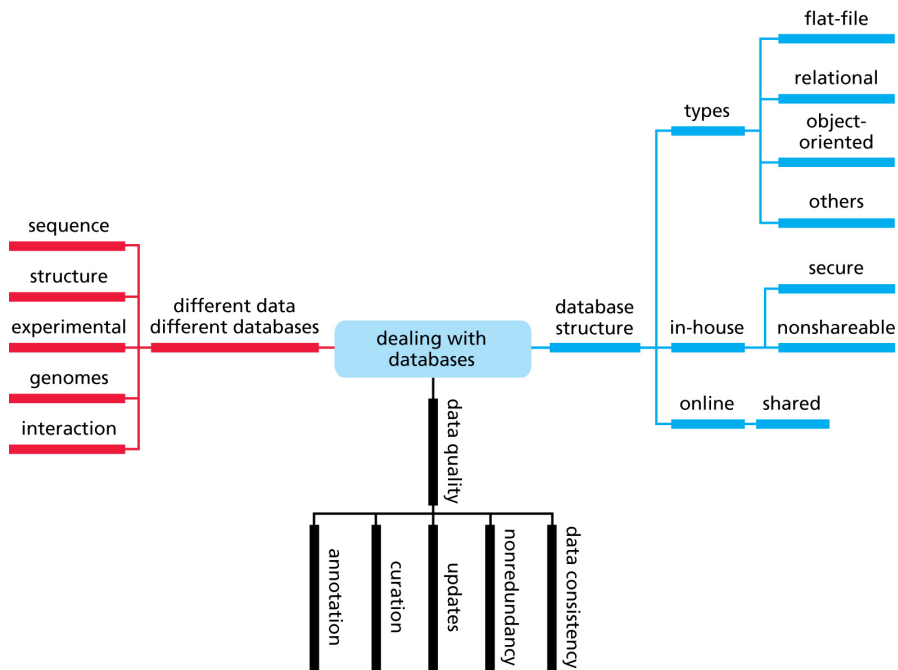
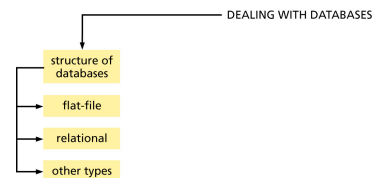


General aspects of databases



Structure of databases



Flat database: it is the simplest form of a database where collections of data (aminoacid sequence) are stored as a large txt file or more than one txt file.

Relational database: it stores the data within a number of tables, each consisting of records and fields. Each table will be linked to at least one other by a shared field called a **KEY**.

protab1			
Protein-code	Protein-name	Length	Species-origin
P1001	Hemoglobin	145	Bovine
P1002	Hemoglobin	136	Ovine
P1003	Eye Lens Protein	234	Human
.....			

protab2	
Protein-code	Protein-sequence
P1001	MDRTTHGFDLKLLSPRTVNQWLMLALFFGHS...
P1002	MDKTSHGFEIKLLTPKKLQQWLMIAIYFGHT...
P1003	SRTHEEEGKLMQWPPRPLYIALFTEPPYP...
.....	

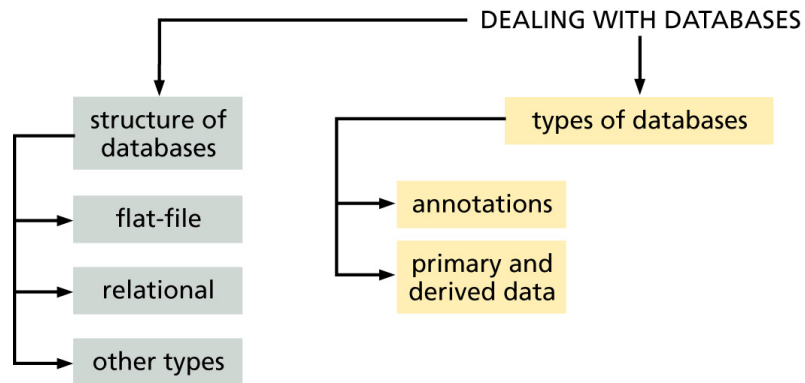
Type of databases

Data: it is the minimal content of a database including data's identity (for example protein name and source) and the author/submitter responsible for the entry.

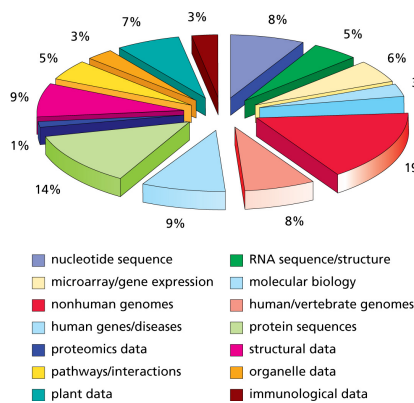
Annotation: provide more information to the data (published papers, lists of entries in other databases, gene structure)

Primary data: they include the raw experimental results.

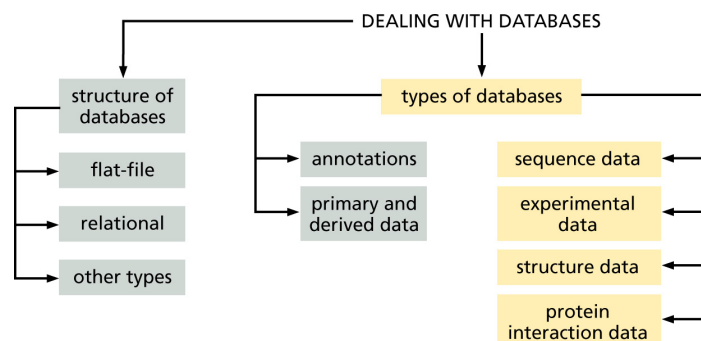
Derived data: based on the data existing at the time (example: conserved protein sequence motifs).



Looking for databases



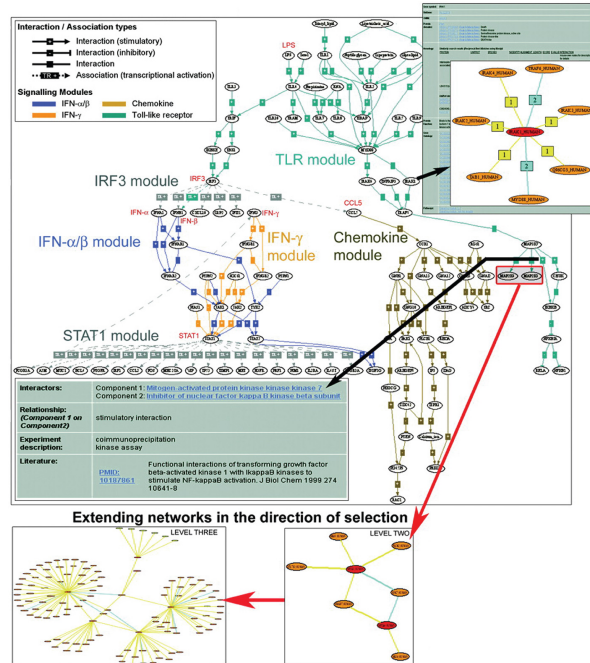
Distribution of the type of databases as classified at the Nucleic Acid Research (NRA) Molecular Biology Database Collection Web site. In 2006 there were 858 databases listed, classified into 14 main categories.



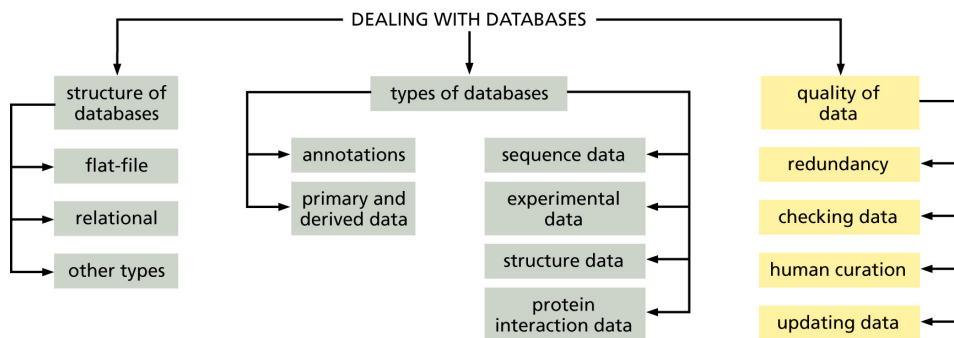
Protein interaction databases

They provide information about the interactions of proteins with other molecules, including other proteins.

They include: the Database of Interacting Proteins (DIP) and the Molecular INteraction Database (MINT).



Quality of databases



Non redundant databases: they include all the experimental data (from different labs) in one entry.

Checking data: a DNA sequence must contain only A, C, G, T. A protein sequence must correspond to a certain molecular weight according to the amino acids present.

Sequence alignments

Similarity: the sequences show some degree of match.

Homology: similarity in sequence or structure due to descent from a common ancestor.

Mutation and selection over millions of years can result in considerable divergence between present-day sequences derived from the same ancestral gene.

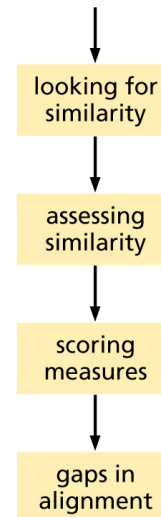
Bases at originally same position can change as a result of:

- Mutations
- Insertions
- Deletions
- Gene fusions

Homology \Rightarrow common ancestor \Rightarrow common structure or function?

Not always.....

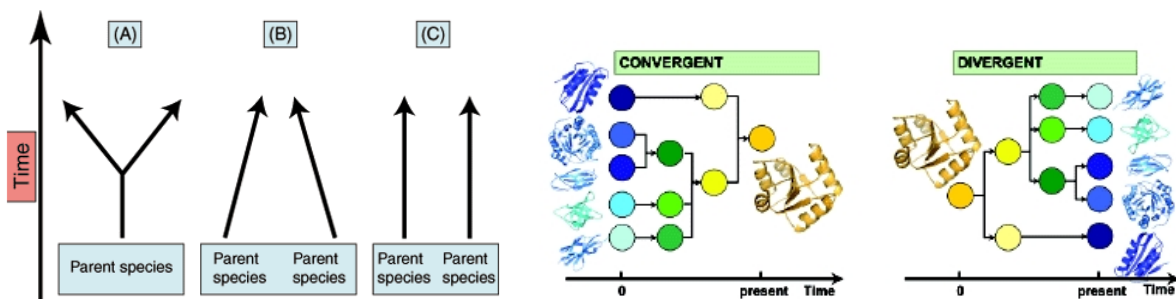
PRODUCING AND ANALYZING SEQUENCE ALIGNMENTS



Sequence alignments

Divergent evolution: mutation and selection can generate proteins with new functions but relatively little changes in sequence. Therefore, sequence similarity does not always imply a common function.

Convergent evolution: proteins with very little sequence similarity to each other but in which a common protein fold and function are preserved.



It is easier to compare to detect homology when comparing protein sequence than when comparing nucleic acid sequences.

1. There are only 4 letters to compare in the DNA alphabet compared to the 20 letters in the protein one
2. The genetic code is redundant
3. The 3D structure of a protein and hence its function, is determined by the amino acid sequence

Scoring alignments

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



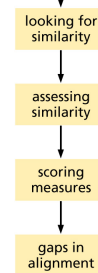
Percent identity: obtained by dividing the number of identical matches by the total length of the aligned region and multiplying by 100.

A good percentage of identity depends on the length of the sequence.



Substitution matrices: the score is assigned to each aligned pair of amino acids by a matrix that defines values for all possible pairs of residues.

PRODUCING AND ANALYZING SEQUENCE ALIGNMENTS

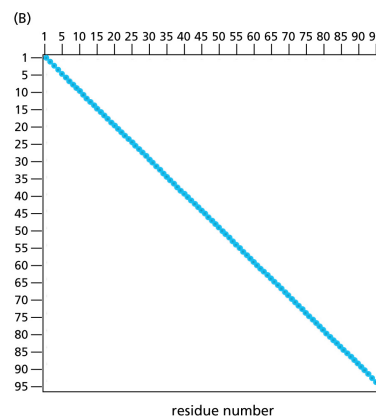
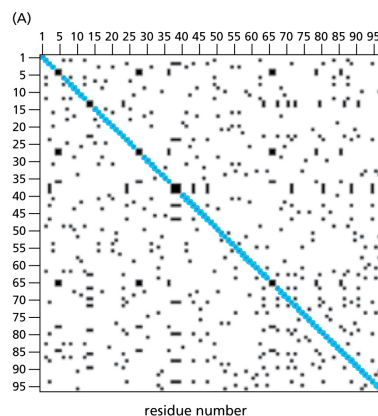
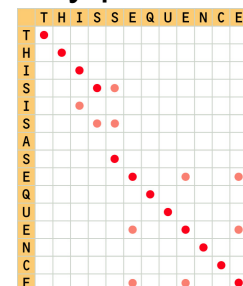


Scoring alignments: identity percentage and similarity percentage

Dot-plots: it is the simplest way to compare sequence similarities.

Use of filters:

- Window size allows to overlap fixed-length windows
- Minimum identity score: it is the minimum identity score fixed for the window previously set.



Two views of dot-plot representations of an SH2 sequence compared to itself. A) Unfiltered dot-plot. The identity is shown by the unbroken diagonale. There is some background noise. B) Dot-plot of the same sequence comparison with a window of 10 residues and a minimum identity score within the window set to 3.

Scoring alignments: identity percentage and similarity percentage

Similarity percentage: it takes into account the so-called conservative substitution

THISISA-SEQUENCE

TH---ATSEQUENCE

THISISA-SEQUENCE
 THAT---SEQUENCE

```

gi|66361410|pdb|1ZBM|A          -----MGHHHHHSHKIRVAHTPDADD 22
gi|154175534|ref|YP_001409022.  -----MKNIKHIDVAHSPDADD 17
gi|6647837|sp|O28098.1|SUCD2_A  -----MAIIVDERTKVVGQITGYQK 22
gi|1711576|sp|P53598.1|SUCA_Ye  MLRSTVSKAKLKI CRHFHRESIPYDKTIKNLLPKDKTIVFQGFQKQGT 50
                                     . . . . .

gi|66361410|pdb|1ZBM|A          AFXYFAXTHGKVDI-WLEIEHVIEDIETLNKRFKAFNAEYEVTAISAHAYAL 71
gi|154175534|ref|YP_001409022.  IFYMAIKFGWVSGKNSLFTALDQLTNEALKSTYATAISFALYPL 67
gi|6647837|sp|O28098.1|SUCD2_A  FHTERMLNYGTKIVAGVTPGKGGTEVLGVPVYDSVKEAVREADANASVIF 72
gi|1711576|sp|P53598.1|SUCA_Ye  FHASISQEYGTNVVGGINPKAGQTHLQGPVFSVKAIDKGTGATASAI 100
                                     *
                                     . . . . .

gi|66361410|pdb|1ZBM|A          LDDKRYILSAGASVGDGYPVVVAKSETSLD-GKRIAVFGRYTTANLLK 120
gi|154175534|ref|YP_001409022.  ISDDYALLRCVAVSFGEGYFKLKKRGNLKRNFVVALSGAHTINALLFR 117
gi|6647837|sp|O28098.1|SUCD2_A  VFAPFAADAVMEADAGIKVIVCITEGIPVHDELKMYRWKREAGAT-LIG 121
gi|1711576|sp|P53598.1|SUCA_Ye  VFPPIAAAATKESIEAETPLAVCTEIGIPQHDMLYIAEMLQTDKTLV 150
                                     . . . . .

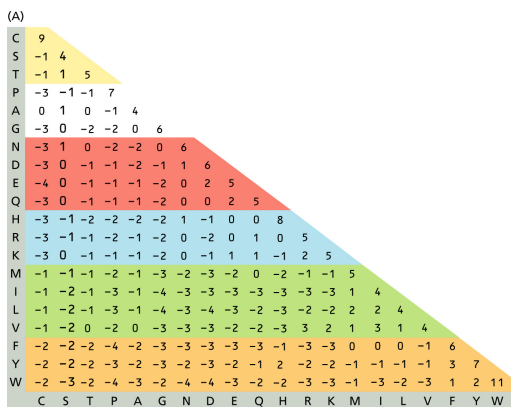
gi|66361410|pdb|1ZBM|A          LAVE-DFEPVEXPFDRIIQAVLDEEVDAGLLIHEGQITYADYGLKCVL 169
gi|154175534|ref|YP_001409022.  AAYP-EARIVYKNFLEIENAVLSGEVDAGVLIHESLGFSS-ELEVEREI 165
gi|6647837|sp|O28098.1|SUCD2_A  PNCPGIISPG-KTHLGMPVQVFKPKNVGVLSRSGTLTYQIAYNLKGL 170
gi|1711576|sp|P53598.1|SUCA_Ye  PNCPGIINPATKVRIGIQPPKIFQAGKIGIISRSGLTYEAQQTKTDL 200
                                     *
                                     . . . . .

gi|66361410|pdb|1ZBM|A          WDWVSEQV--KLPLPLGNAIRRDLSEVQEEFLRAXRESIAFAIEN-PD 216
gi|154175534|ref|YP_001409022.  WDWVCELAGEMLPLPLGGMALRRSLPLDAIECERVLTKAVAIATAHKPF 215
gi|6647837|sp|O28098.1|SUCD2_A  GQSTVVGIGGDRITGDFVEVLRLEFDDKTEKAVVVLVGEIGGRDEEVAE 220
gi|1711576|sp|P53598.1|SUCA_Ye  GQSLVIGMGDFAFPPTDFIDALKLFLEDETTETGIIIMLGEIKAEIAEA 250
                                     . . . . .

gi|66361410|pdb|1ZBM|A          EAIYAX-----KYSRGLDRERAKRFAXXYVNDYTYNXFESVDAAL 257
gi|154175534|ref|YP_001409022.  LSHMLME-----RNLIRIDKELKIYLNLYANKDSISMNETQLKAL 256
gi|6647837|sp|O28098.1|SUCD2_A  FIREMS-----KPVVGVAGLTAPPGR--RMGHAGATIEGGVGTAEISKI 262
gi|1711576|sp|P53598.1|SUCA_Ye  FLKRYNFSRSKMPFVASFIAGTVAGQMKGVYRHHSGAIVGSGGTDAESKK 300
                                     . . . . .

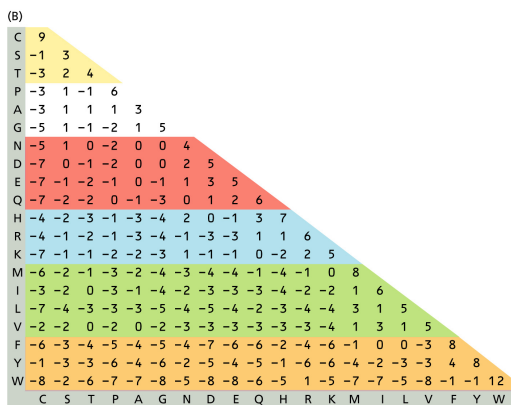
gi|66361410|pdb|1ZBM|A          KKLVEY-----AEANGLIKMPKLDILRL-- 280
gi|154175534|ref|YP_001409022.  NRLFIEGYDQGFYPOPIDAHDVLIPTVEVNDARFS- 290
gi|6647837|sp|O28098.1|SUCD2_A  KALEAAG-----ARVGTPEVVAELVLAELL-- 287
gi|1711576|sp|P53598.1|SUCA_Ye  QALRDVG-----VAVVESPGVLQALLDQFAFK 329
                                     : *
                                     . . . . .
    
```

Scoring alignments: substitution matrices



Expectation value (E-value): the probability of different alignments with scores equivalent to or better than S that are expected to occur in a database search by chance. The lower the E value, the more significant the score.

It indicates the number of sequences that would be expected to have that score (or more) if the query sequence were compared against a database containing no sequences related to the query sequence. Thus, a lower E-value indicates that the sequences are more likely to be related than if the comparison had a higher E-value. An E-value of 0.00001 or less (also sometimes written as $1e-5$, which is shorthand for 1.0×10^{-5}) is often used as good initial evidence that a query and database sequence are related, although further investigation should always be carried out to obtain additional support for such a hypothesis.



Amino acids substitution scoring matrices. A) The BLOSUM-62 matrix and B) the PAM120 matrix. The colored shading indicates different physicochemical properties of the residues.

Sequence alignments: BLAST



results of **BLAST**

Description	BLAST equivalent
<p>Protein compared to protein database or DNA to DNA database. For protein, ktup = 2 by default (ktup = 1 is more sensitive); default for DNA is 6; 4 or 3 is more sensitive. 1 should be used for short DNA stretches.</p> <p>Uses Smith-Waterman algorithm. Can search protein to protein or DNA to DNA. Can be more sensitive than fasta with protein sequences.</p>	blastp/blastn
<p>DNA compared to protein database. DNA translated into all three frames. fastx slower than fasty but better. Used to see if DNA encodes a protein.</p>	blastx
<p>Protein compared to DNA database. Mainly used to identify EST sequences. This is preferred over fastx as protein comparison is more sensitive than DNA.</p>	tblastn (tblastx compares translated DNA to translated DNA database)
<p>Mixed peptide sequence (such as obtained by Edman degradation) compared to protein database.</p>	
<p>Mixed peptide sequence compared to DNA database.</p>	

(A)

sp P32871 P11A	BOVIN	PHOSPHATIDYLINOSITOL 3-KINASE CATALYT...	680	0.0
sp P42336 P11A	HUMAN	PHOSPHATIDYLINOSITOL 3-KINASE CATALYT...	676	0.0
sp P42337 P11A	MOUSE	PHOSPHATIDYLINOSITOL 3-KINASE CATALYT...	674	0.0
sp P42338 P11B	HUMAN	PHOSPHATIDYLINOSITOL 3-KINASE CATALYT...	338	9e-93
sp O35904 P11D	MOUSE	PHOSPHATIDYLINOSITOL 3-KINASE CATALYT...	332	7e-91
sp O00329 P11D	HUMAN	PHOSPHATIDYLINOSITOL 3-KINASE CATALYT...	331	2e-90
↓				
sp P47473 R1R1	MYCGE	RIBONUCLEOSIDE 2-DIPHOSPHATE REDUCTASE A...	34	0.59

(B) **Distribution of 2 Blast Hits on the Query Sequence**

Mouse-over to show define and scores. Click to show alignments

(C)

- This CD alignment includes 3D structure. To display structure, download Cn3D v3.00!

Mouse-over boxes to display more information

Sequences producing significant alignments:

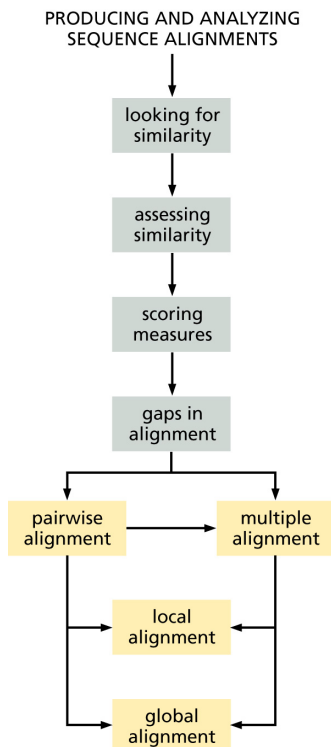
Accession	Description	Score	E value
gnl Smart P13Kc	Phosphoinositide 3-kinase, catalytic domain, Phosphoinositide ...	201	3e-83
gnl Pfam dm 0054 P13_P14_kinase	Phosphatidylinositol 3- and 4-kinases	263	9e-72

ADD query to multiple alignment, display up to 10 sequences [most similar to the query]

Length = 265
Score = 301 bits (763), Expect = 3e-83

Query: 19 IFFNDDGLRQDMITGQIFIRKINIQWQGLLIRNLPVGLRIGDCVGLLEVERNSMIR 78
Subject: 2 IFFNDDGLRQDMITGQIFIRKINIQWQGLLIRNLPVGLRIGDCVGLLEVERNSMIR 61

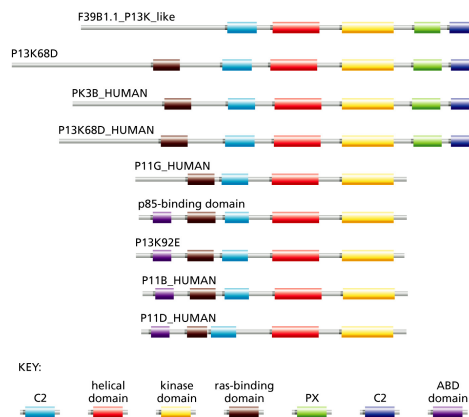
Types of alignments



Global alignment: it is used to find or compare closely related sequences that are similar over their whole sequence.

Local alignment: can reveal that parts of sequences are related.

It is useful in multidomain proteins.



P13-kinase is a multidomain protein. Output from Pfam.

Multiple alignments

They can be constructed by different techniques.

(A) structural/functional alignment from BALiBase

```

1csy SHEKMPWFHGKISRESEQIVLIGSKTNGKFLIRARD--NNGSYALCLLHEGKVLHYRIDKDKTGKLSIPEGK-KFDTLWQLVEHYKYA-----DGLLRVLT-TVPCKK
1gr1 EMKPHWFFGKIPAKAEML-SKQRHOGAFLIRESES-APGDFLSVSKFGNDVQHFKVLRDAGKGYFLWVW-KFNLSNELVDYHRST-SVSRNQQIFLRDIEQVPAQ-
1aya ---MRRWFHPNITGVEAENLLLRG-VDGSFLARPSKS-NPGDFTLSVRRNGAVTHIKIGN--TGDDYDLYGGEKFATLAELVQYMEHHGQLEKNGDVIEL-KYPLN-
2pna -LQDAEWMYGDISREEVNEKLRDT--ADGTFLVRDASTKMHGDYTLRLKGGNNKLIKIFHR-RDGKYGFSPLT-FNSVVELINHYRNE-SLAQYMPKLDVKL-LYPVS-
1bf1 HHDEKTNWVGSSRNKAENLLRGK--RDGTFLVRES--KRGCYACSVVDGVEKHCVINKTATGYGFAEPYNYSSLKEVLVHYQHT-SLVQHNSLVNLT-LYFYVA
    
```

(B) DIALIGN multiple sequence alignment

```

1csy SHEKMPWFHGKISRESEQIVLIGSKT-NGKFLIRAR-DN--NNGSYALCLLHEGKVLHYRIDKDKTGKLSIPEGK-KFDTLWQLVEHYKYA-----DGLLRVLT-TVPCKK
1gr1 EMKPHWFFGKIPAKAEML--SKQRHOGAFLIRESES--PGDFLSVSKFGNDVQHFKVLRDAGKGYFLWVW-K-FNLSNELVDYHRST--SVSRNQQIFLRDIEQVPAQ-
1aya M---MRRWFHPNITGVEAENLLLRGV--DGSFLARPSKSN--PGDFTLSVRRNGAVTHIKIGN--TGDDYDLYGGEKFATLAELVQYMEHHGQLEKNGDVIEL-KYPLN-
2pna LQDAE-WYMGDISREEVNEKLL--RDTA-DGTFLVRDA-STKMHGDTYTLRLKGGNNKLIKIFHRDGKYGFSO-PLT-FNSVVELINHYRNE--SLAQYMPKLDVKLL-LYPVS-
1bf1 HHDEKTNWVGSSRNKAENLL--RGRK-DGTFLVRES-SK--QGCYACSVVDGVEKHCVINKTATGYGFAE-PYNYSSLKEVLVHYQHT--SLVQHNSLVNLT-LYFYVA
    
```

(C) ClustalW multiple sequence alignment

```

1csy SHEKMPWFHGKISRESEQIVLIGSKTNGKFLIRARDN--NNGSYALCLLHEGKVLHYRIDKDKTGKLSIPEGK-KFDTLWQLVEHYKYA-----ADGLLRVLT-TVPCKK
1gr1 EMKPHWFFGKIPAKAEML-SKQRHOGAFLIRESES-APGDFLSVSKFGNDVQHFKVLRDAGKGYFLWVW-K-FNLSNELVDYHRST-SVSRNQQIFLRDIEQVPAQ-
1aya ---MRRWFHPNITGVEAEN-LLLRGVDGSLARPSKS-NPGDFTLSVRRNGAVTHIKIGN--TGDDYDLYGGEKFATLAELVQYMEHHGQLEKNGDVIEL-KYPLN-
2pna -LQDAEWMYGDISREEVN--EKLRTADGTFLVRDASTKMHGDYTLRLKGGNNKLIKIFHR-DGKYGFSPLT-FNSVVELINHYRNE-SLAQYMPKLDVKLL-LYPVS-
1bf1 HHDEKTNWVGSSRNKAEL--NLLRGRDGTFLVRESK--RGCYACSVVDGVEKHCVINKTATGYGFAEPYNYSSLKEVLVHYQHT-SLVQHNSLVNLT-LYFYVA
    
```

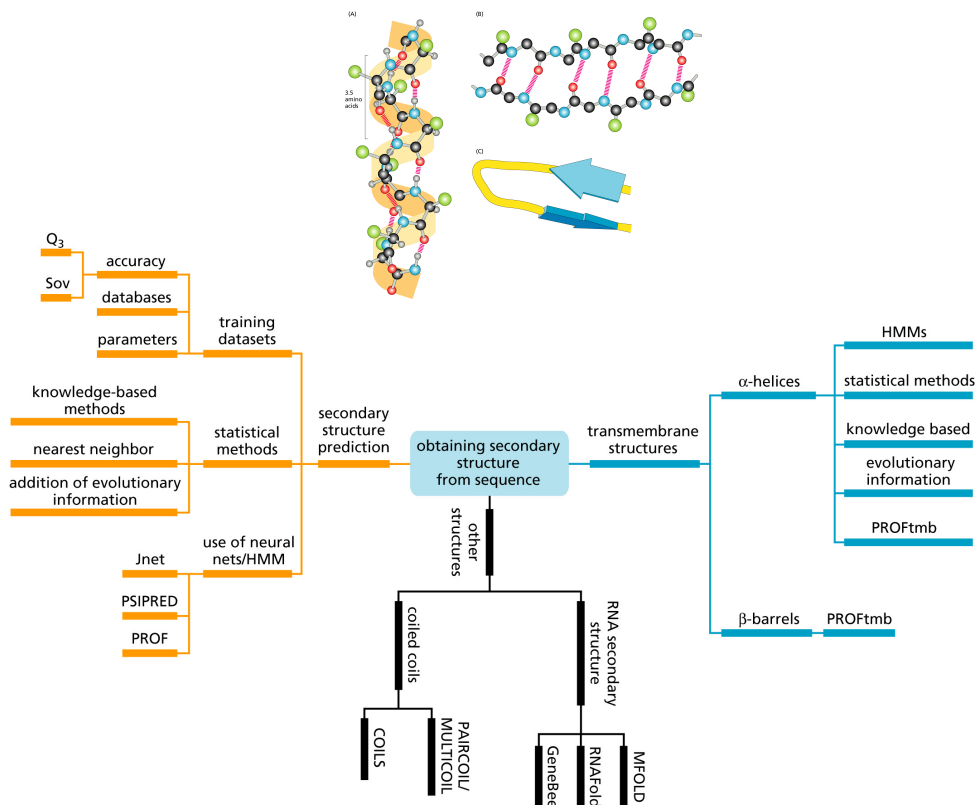
(D) divide-and-conquer multiple sequence alignment

```

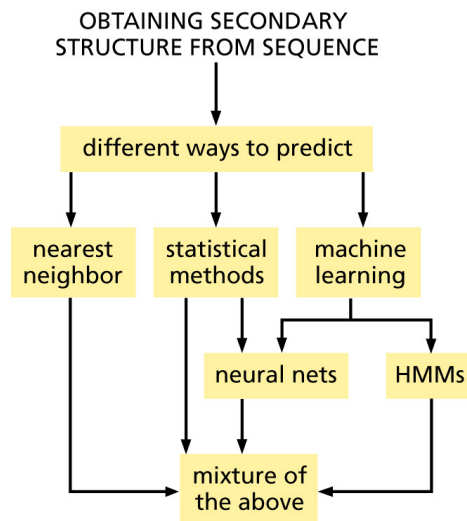
1csy SHEKMPWFHGKISRESEQIVLIGSKTNGKFLIRARDN--NNGSYALCLLHEGKVLHYRIDKDKTGKLSIPEGK-KFDTLWQLVEHYKYA-----KADGLLRV-L-TVPCKK
1gr1 EMKPHWFFGKIPAKAEML-SKQRHOGAFLIRESES-APGDFLSVSKFGNDVQHFKVLRDAGKGYFLWVW-K-FNLSNELVDYHRST-SVSRNQQIFLRDIEQVPAQ-
1aya ---MRRWFHPNITGVEAENLLLRGVDGSLARPSKS-NPGDFTLSVRRNGAVTHIKIGN--TGDDYDLYGGEKFATLAELVQYMEHHGQLEKNGDVIEL-KYPLN-
2pna -LQDAEWMYGDISREEVN--RDTADGTFLVRDASTKMHGDYTLRLKGGNNKLIKIFHRDGKYGFSO-PLT-FNSVVELINHYRNE-SLAQYMPKLDVKLL-LYPVS-
1bf1 HHDEKTNWVGSSRNKAENLL--RGRDGTFLVRESK--RGCYACSVVDGVEKHCVINKTATGYGFAEPYNYSSLKEVLVHYQHT-SLVQHNSLVNLT-LYFYVA
    
```

Structural alignments: if the structure of one of the proteins is known, then the gap penalty can be increased for regions of known secondary structure such as helices and strands, as these regions are less likely to suffer insertions or deletions. This will mean that few or no gaps are introduced into these regions.

Protein secondary structure prediction



Types of secondary structure prediction



Statistical methods are based on rules that give the probability that a residue will form part of a particular secondary structure.

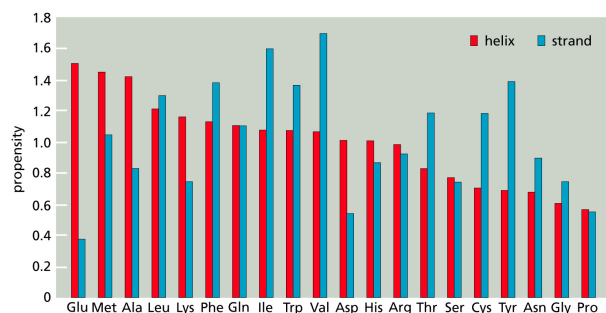
The probabilities are derived from analysing structure and sequence data from large sets of proteins of known structure.

Nearest neighbor methods are statistical methods that incorporate additional information about protein structure (shapes, sizes and physicochemical properties of the different amino acid residues).

Machine learning approaches train a neural net or other learning algorithms to acquire structure-sequence relationships which can then be applied to predict structure from a protein sequence.

Statistical and knowledge-based methods: Chou and Fasman

A.A.	P(a)	P(b)	P(turn)	f(i)	f(i+1)	f(i+2)	f(i+3)
Alanine	142	83	66	0.060	0.076	0.035	0.058
Arginine	98	93	95	0.070	0.106	0.099	0.085
Asparagine	67	89	156	0.161	0.083	0.191	0.091
Aspartic acid	101	54	146	0.147	0.110	0.179	0.081
Cysteine	70	119	119	0.149	0.050	0.117	0.128
Glutamic acid	151	37	74	0.056	0.060	0.077	0.064
Glutamine	111	110	98	0.074	0.098	0.037	0.098
Glycine	57	75	156	0.102	0.085	0.190	0.152
Histidine	100	87	95	0.140	0.047	0.093	0.054
Isoleucine	108	160	47	0.043	0.034	0.013	0.056
Leucine	121	130	59	0.061	0.025	0.036	0.070
Lysine	114	74	101	0.055	0.115	0.072	0.095
Methionine	145	105	60	0.068	0.082	0.014	0.055
Phenylalanine	113	138	60	0.059	0.041	0.065	0.065
Proline	57	55	152	0.102	0.301	0.034	0.068
Serine	77	75	143	0.120	0.139	0.125	0.106
Threonine	83	119	96	0.086	0.108	0.065	0.079
Tryptophan	108	137	96	0.077	0.013	0.064	0.167
Tyrosine	69	147	114	0.082	0.065	0.114	0.125
Valine	106	170	50	0.062	0.048	0.028	0.053



Chou-Fasman is one of most commonly used algorithms

- measured frequencies at which each amino acid appeared in particular types of secondary sequences in a set of proteins of known structure
- assigns the amino acids three conformational parameters based on the frequency at which they were observed in alpha helices, beta sheets and beta turns
 1. **P(a) = propensity to form alpha helices**
 2. **P(b) = propensity to form beta sheets**
 3. **P(turn) = propensity to form beta turns**
- also assigns 4 turn parameters based on frequency at which they were observed in the first, second, third or fourth position of a beta turn
 1. **f(i) = probability of being in position 1**
 2. **f(i+1) = probability of being in position 2**
 3. **f(i+2) = probability of being in position 3**
 4. **f(i+3) = probability of being in position 4**

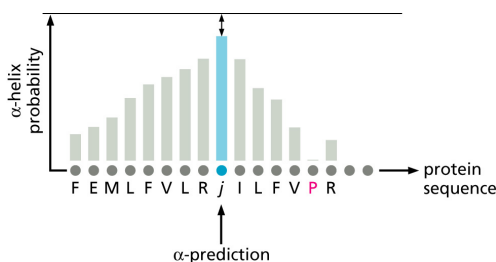
Statistical and knowledge-based methods: Chou and Fasman

identifies helix and sheet "nuclei", then applies a set of heuristic rules to determine if these clusters of amino acids are sufficient to nucleate a region of alpha-helix or beta-sheet.

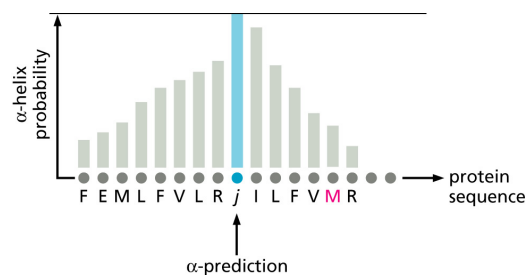
- **helix: 4 out of 6 amino acids with $P(a) > 100$**
 - extends the nucleus in each direction until reach four amino acids in a row with $P(a) < 100$
 - for each of these regions, add up all the $P(a)$ and all the $P(b)$ values.
 - If the total $P(a)$ is larger than the total of $P(b)$ and the run is more than 5 amino acids long, then it is predicted to be alpha helix
- **sheet: 4 out of 6 amino acids with $P(b) > 100$ (some people use 3 out of 5).**
 - extends the nucleus in each direction until reach four amino acids in a row with $P(b) < 100$
 - for each of these regions, add up all the $P(a)$ and all the $P(b)$ values.
 - If the total $P(b)$ is larger than the total of $P(a)$, the run is more than 5 amino acids long, and the average $P(b) > 100$ then it is predicted to be beta sheet.
- **If helices and sheets overlap then compare the total $P(a)$ and total $P(b)$ for the overlapping region. If the total $P(a)$ is larger than the total of $P(b)$ then it is predicted to be alpha helix (and vice-versa)**
- **beta turn**
 - calculate the likelihood of a turn $P(t)$ for amino acid at position i as the sum of $f(i) +$ the $f(i+1)$ value for the following amino acid + the $f(i+2)$ value for the next amino acid + the $f(i+3)$ value for the amino acid at the plus three position.
 - Predict a beta- turn at position i if the following criteria are met:
 - the calculated $P(t)$ is > 0.5
 - the average $P(\text{turn})$ for amino acids i to $i+3$ is > 100
 - the sum of the $P(\text{turn})$ values for amino acids i to $i+3$ is larger than the sum of the $P(a)$ and $P(b)$ values
- Accuracy = 50-85%, depending on the protein

Statistical and knowledge-based methods: GOR

It incorporates the effects of local interactions between amino acids residues by taking successive windows of 17 residues and considering the effect of residues from position $j-8$ to $j+8$ on the conformation of the residue at position j .



The effect of an helix breaker (Pro) at position $j+5$. The proline diminishes the overall additive propensity of residue j to form helix



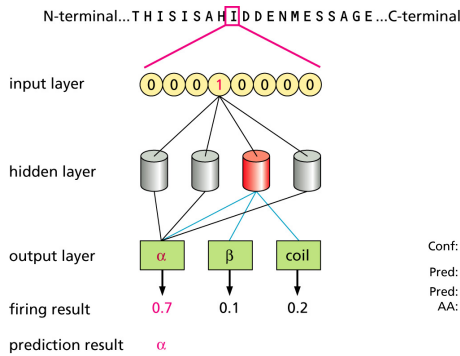
The effect of a non helix breaker (Met) at position $j+5$. The methionine improves the overall additive propensity of residue j to form helix

Neural networks methods

The algorithm will learn by iterative changes to its parameters until the predicted structure is as similar to the observed structure as possible.

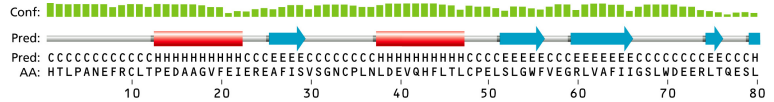
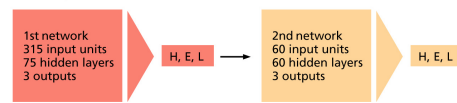
PSIPRED is a three stage method:

1. It generates a multiple sequence alignment
2. It generates an initial secondary structure
3. It filters the initial prediction

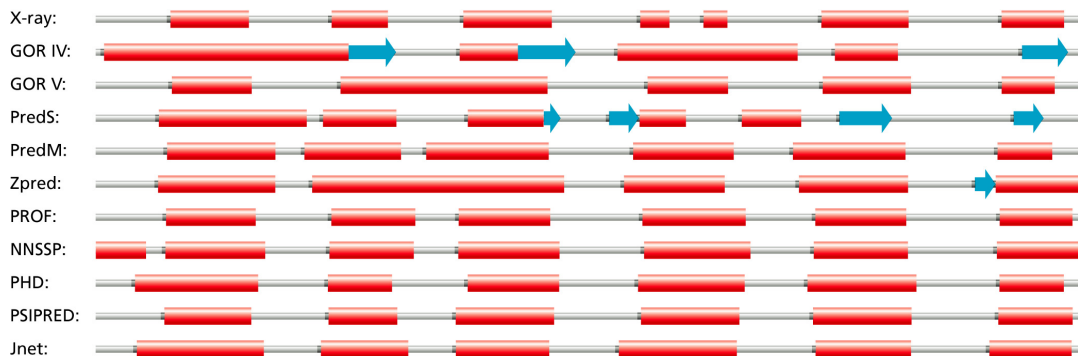
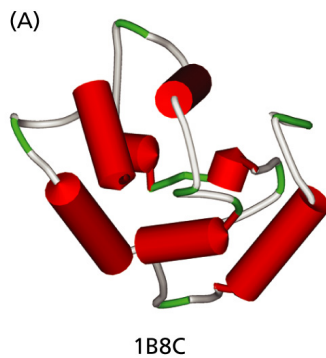


A	R	N	D	C	Q	E	G	H	I	L	K	M	F	P	S	T	W	Y	V
-3	-4	-4	-4	-3	-4	-4	-4	-2	-1	-1	-4	-1	8	-5	-3	-3	0	2	-2
0	-1	-1	3	-4	3	4	1	-1	-4	-4	0	-3	-4	-2	-1	-2	-4	-3	-3
0	-1	2	1	-3	4	0	-1	-2	-4	-3	1	-2	-4	-2	2	0	-4	-3	-3
-2	-3	-4	-5	-2	-3	-4	-6	-4	0	6	0	0	-1	4	-2	2	0	-4	-2
0	-3	-1	-2	0	-2	4	-3	-3	0	-2	-4	-3	3	1	-4	-4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

A	R	N	D	C	Q	E	G	H	I	L	K	M	F	P	S	T	W	Y	V
0.4	0.3	0.3	0.3	0.2	0.9	0.3	0.3	0.4	0.4	0.4	0.4	0.4	0.9	0.1	0.4	0.4	0.5	0.7	0.4
0.3	0.2	0.3	0.8	0.4	0.3	0.7	0.1	0.6	0.2	0.3	0.3	0.5	0.2	0.1	0.4	0.8	0.2	0.3	0.2
0.1	0.1	0.4	0.3	0.5	0.1	0.1	0.3	0.1	0.1	0.4	0.2	0.4	0.9	0.3	0.4	0.4	0.9	0.3	0.6
.

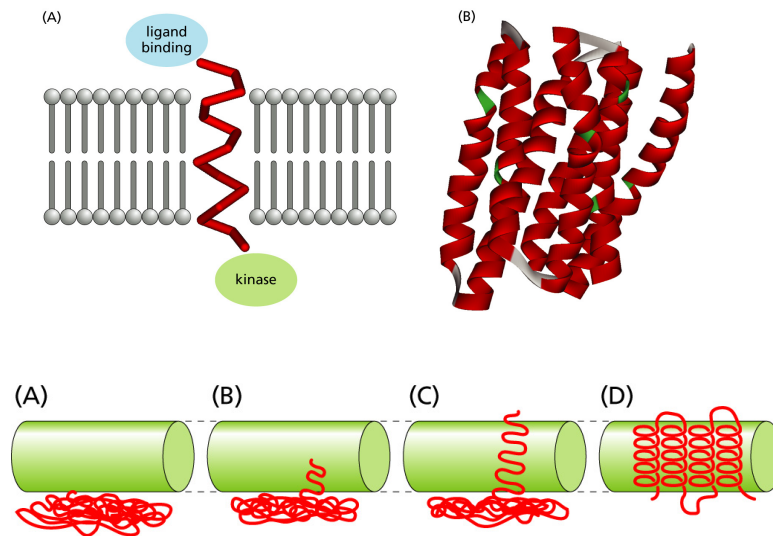


Secondary structure prediction methods



Transmembrane proteins

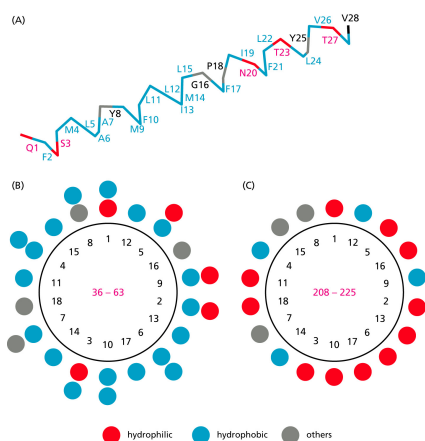
Membrane proteins are functionally important. For example, the receptors are formed by 1 or more helices spanning the membrane



The four main ways in which proteins may be attached to a membrane. A) Attachment by interactions between the protein and the cytosolic face of the lipid bilayer. B) Attachment via an anchor (lipidic or terminals of the protein) that are added post-translationally. C) Transmembrane proteins have part of the protein chain embedded in the lipid bilayer. D) Transmembrane proteins where the protein chain threads back and forth across the membrane multiple times.

Transmembrane proteins

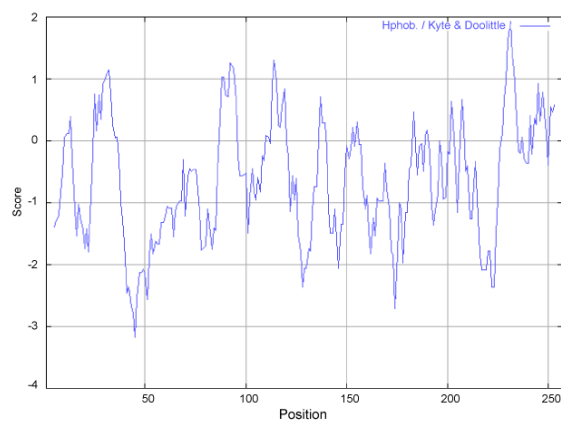
Helix wheel



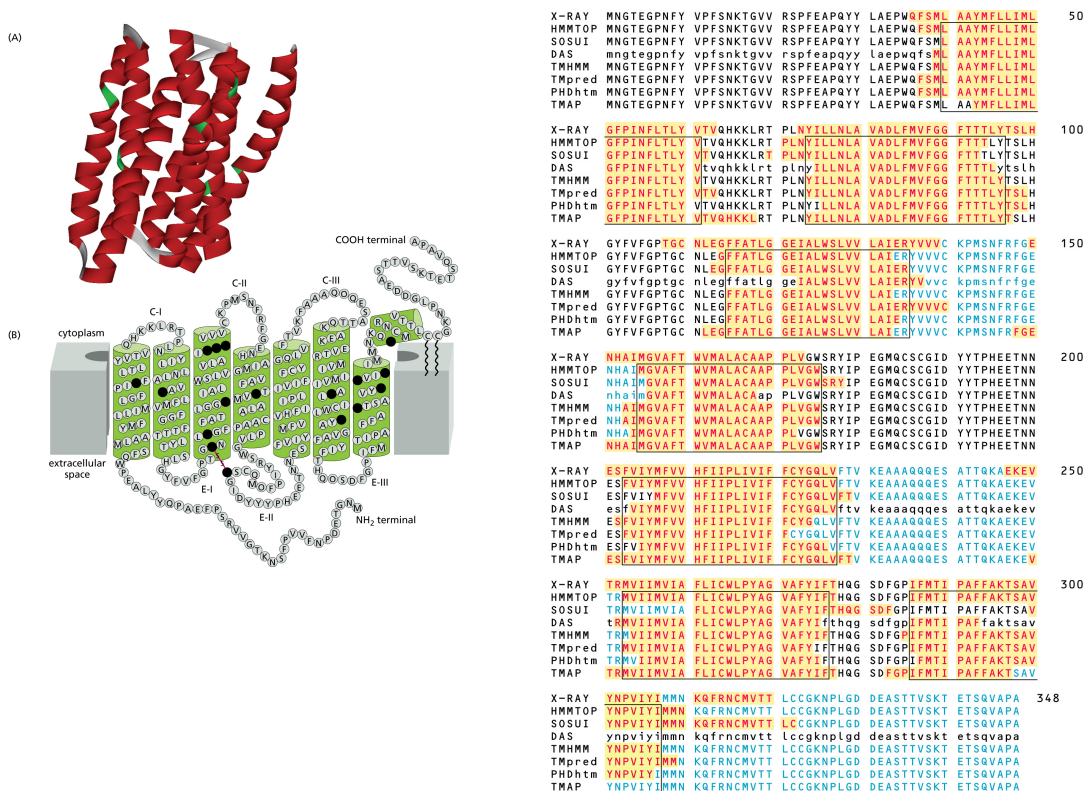
Hydrophobicity diagram

Using the scale [Hphob. / Kyte & Doolittle](#), the individual values for the 20 amino acids are:

Ala: 1.800	Arg: -4.500	Asn: -3.500	Asp: -3.500	Cys: 2.500	Gln: -3.500
Glu: -3.500	Gly: -0.400	His: -3.200	Ile: 4.500	Leu: 3.800	Lys: -3.900
Met: 1.900	Phe: 2.800	Pro: -1.600	Ser: -0.800	Thr: -0.700	Trp: -0.900
Tyr: -1.300	Val: 4.200	: -3.500	: -3.500	: -0.490	



Transmembrane proteins



Protein Sequence Motifs or Patterns

What is required is a method of searching for the occurrence of short sequence patterns, or motifs.

A motif, in general, is any conserved element of a sequence alignment (CONSENSUS), whether composed of a short sequence of contiguous residues or a more distributed pattern. Functionally related sequences will share similar distribution patterns of critical functional residues that are not necessarily contiguous.



Figure 4.15

Residues that contribute to one of the blocks returned by the BLOCKS database after submission of the PI3-kinase p100 α sequence.

(A) A block for four homologous sequences, and (B) for 31 homologous sequences. These representations are called logos, and are computed using a position-specific scoring matrix. This block contains the active-site amino acids and the DFG kinase motif. The size of the letters indicates the level of conservation and the colors indicate physicochemical properties of the residues: acidic, red; basic, blue; small and polar, white; asparagine and glutamine, green; sulfur-containing amino acids, yellow; hydrophobic, black; proline, purple; glycine, gray; aromatic, orange.

Protein Sequence Motifs or Patterns

The PROSITE database is a compilation of motifs and patterns extracted from protein sequences and compiled by inspection of protein families. This database can be searched with an unknown protein sequence to obtain a list of hits to possible patterns or protein signatures.



Database of protein domains, families and functional sites

[Home](#) [ScanProsite](#) [ProRule](#) [Documents](#) [Downloads](#) [Links](#) [Funding](#)

PROSITE consists of [documentation entries](#) describing protein domains, families and functional sites as well as associated patterns and profiles to identify them [[More details](#) / [References](#) / [Disclaimer](#) / [Commercial users](#)]. PROSITE is complemented by [ProRule](#), a collection of rules based on profiles and patterns, which increases the discriminatory power of profiles and patterns by providing additional information about functionally and/or structurally critical amino acids [[More details](#)].

Release 20.67, of 03-Nov-2010 (1598 documentation entries, 1308 patterns, 909 profiles and 898 ProRule)

PROSITE access

Search:

add wildcard ^{***}

Browse:

- [by documentation entry](#)
- [by ProRule description](#)
- [by taxonomic scope](#)
- [by number of positive hit](#)

PROSITE tools

Scan a sequence against PROSITE patterns and profiles - quick scan
(Output includes graphical view and feature detection)

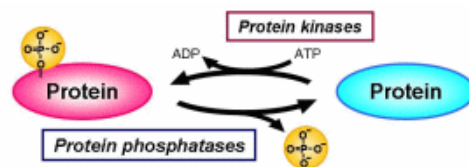
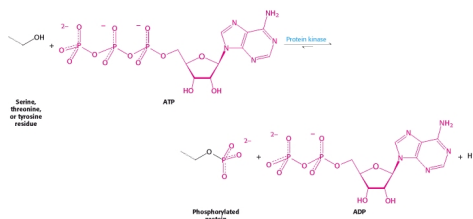
Enter your sequence or a UniProtKB (Swiss-Prot or TrEMBL) ID or AC [[help](#)]:

Protein Sequence Motifs or Patterns

Common covalent modifications of protein activity

Modification	Donor molecule	Example of modified protein	Protein function
Phosphorylation	ATP	Glycogen phosphorylase	Glucose homeostasis; energy transduction
Acetylation	Acetyl CoA	Histones	DNA packing; transcription
Myristoylation	Myristoyl CoA	Src	Signal transduction
ADP-ribosylation	NAD	RNA polymerase	Transcription
Farnesylation	Farnesyl pyrophosphate	Ras	Signal transduction
γ-Carboxylation	HCO ₃ ⁻	Thrombin	Blood clotting
Sulfation	3'-Phosphoadenosine-5'-phosphosulfate	Fibrinogen	Blood-clot formation
Ubiquitination	Ubiquitin	Cyclin	Control of cell cycle

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The consensus sequence recognized by protein kinase A is Arg-Arg-X-Ser-Z or Arg-Arg-X-Thr-Z, in which X is a small residue, Z is a large hydrophobic one, and Ser or Thr is the site of phosphorylation. It should be noted that this sequence is not absolutely required.

Protein Sequence Motifs or Patterns

NetPhos predicts phosphorylation sites in a protein sequence due to kinase acting post-translationally.

```

Name: test1          Length: 26  <-- Sequence name, length
QMERRRTYELVISLIVESYEAYEAY  <-- Submitted sequence
.....T.....SY...Y...      <-- Assignments. S,T,Y indicates

                                predicted phosphorylation sites

Ser: 1 Thr: 1 Tyr: 2          <-- No. of predicted S,T,Y phosph. sites

                                Serine predictions
Name      Pos  Context  Score  Pred
-----
test1     13  ELVISLIVE 0.017  .
test1     18  LIVESYEAY 0.942  *S*

                                Threonine predictions
Name      Pos  Context  Score  Pred
-----
test1     7   ERRRTYELV 0.921  *T*

                                Tyrosine predictions
Name      Pos  Context  Score  Pred
-----
test1     8   RRRTYELVI 0.056  .
test1     19  IVESYEAYH 0.502  *Y*
test1     23  YEAYEAYH- 0.885  *Y*
    
```

