General aspects of databases

Structure of databases

structure of databases
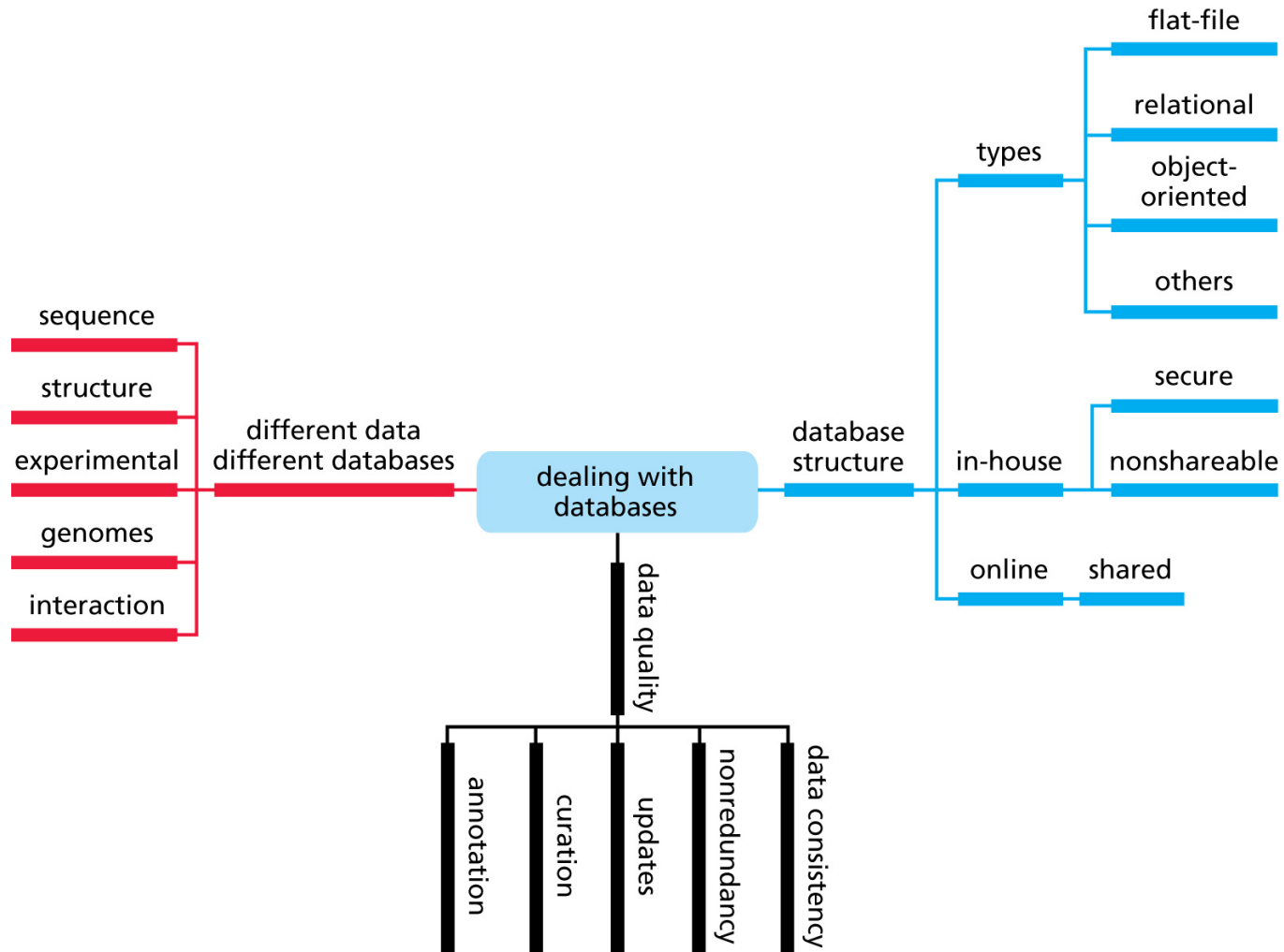
flat-file

relational

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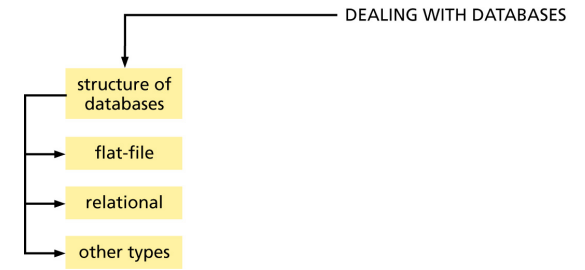
Flat database: it is the simplest form of a database where collections of data (aminoacid sequence) are stored as a large txt file or more than one txt file.

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

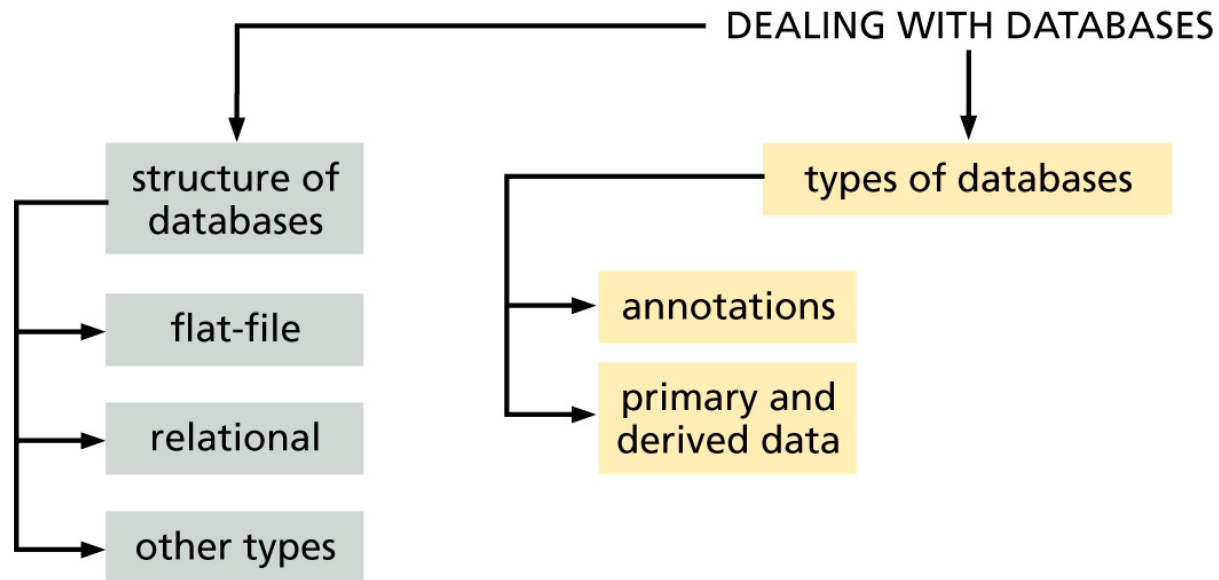| Protein-code	Protein-sequence
P1001	MDRTTHGFDLKLLSPRTVNQWLMLALFFGHS...
P1002	MDKTSHGFEIKLLTPKKLQQWLMIAIYFGHT...
P1003	SRTHEEEGKLMQWPPRPLYIALFTEPPYP...
.....	

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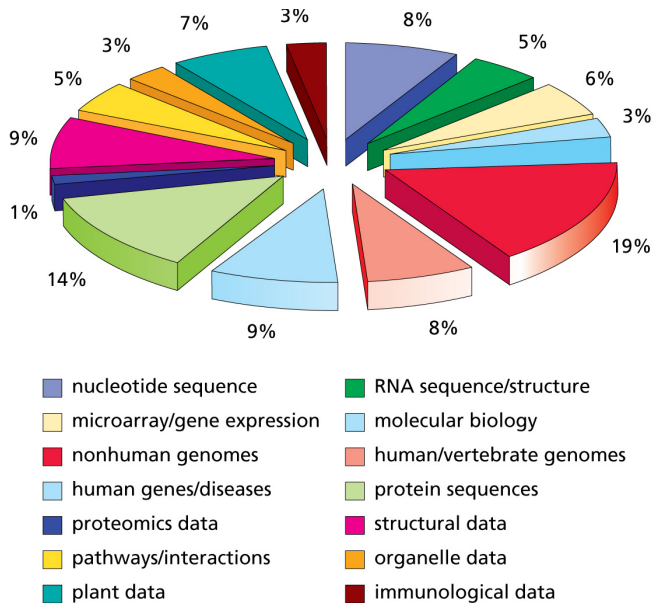
Annotation: provide more information to the data (published papers, lists of entries in other databases, gene structure)

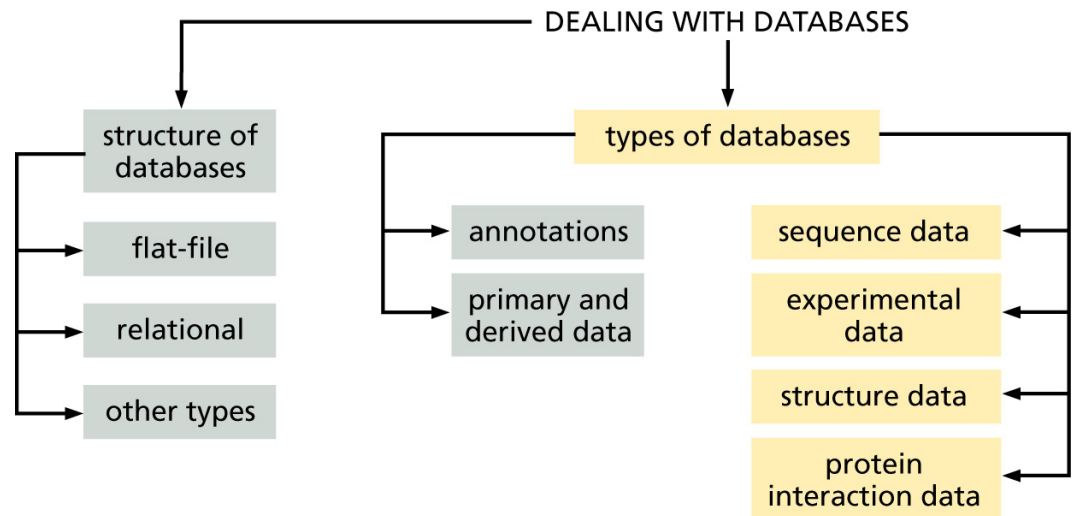***Primary data***: they include the raw experimental results.

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

Looking for databases

- nucleotide sequence
- RNA sequence/structure
- microarray/gene expression
- molecular biology
- nonhuman genomes
- human/vertebrate genomes
- human genes/diseases
- protein sequences
- proteomics data
- structural data
- pathways/interactions
- organelle data
- plant data
- immunological data

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 catagories.

DEALING WITH DATABASES

structure of databases
- flat-file
- relational
- other types

types of databases
- annotations
- primary and derived data
- sequence data
- experimental data
- structure data
- protein interaction data

Sequence database

1. DNA sequences:

- Raw genomic sequence (chromosomal DNA)

- cDNA (from mRNA)

- Expressed sequence tags (ESTs). Partial cDNA seqeunce.

2. Protein sequences (UniProtKB, Swiss-Prot, NCBI Protein Database

Structural database

They contain information about the structure of small molecules, proteins, DNA and RNA sequences, carbohydrates.

Protein folds have also been classified according to the conservation of the fold. They include CATH and SCOP.

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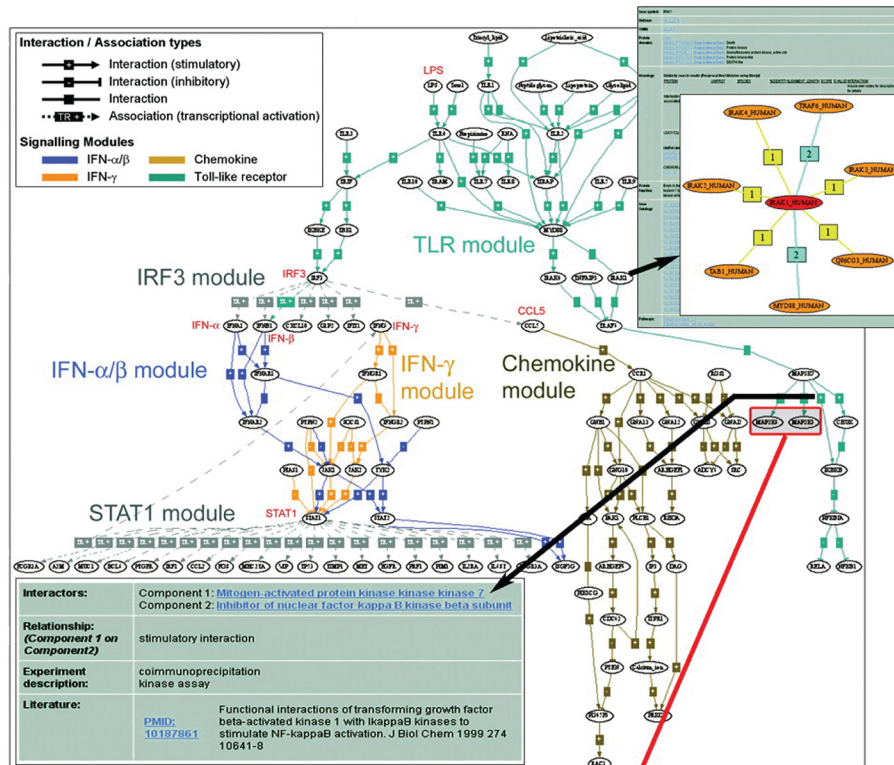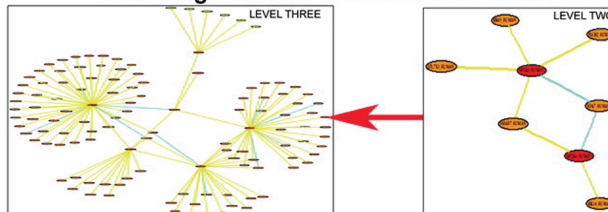
Quality of databases
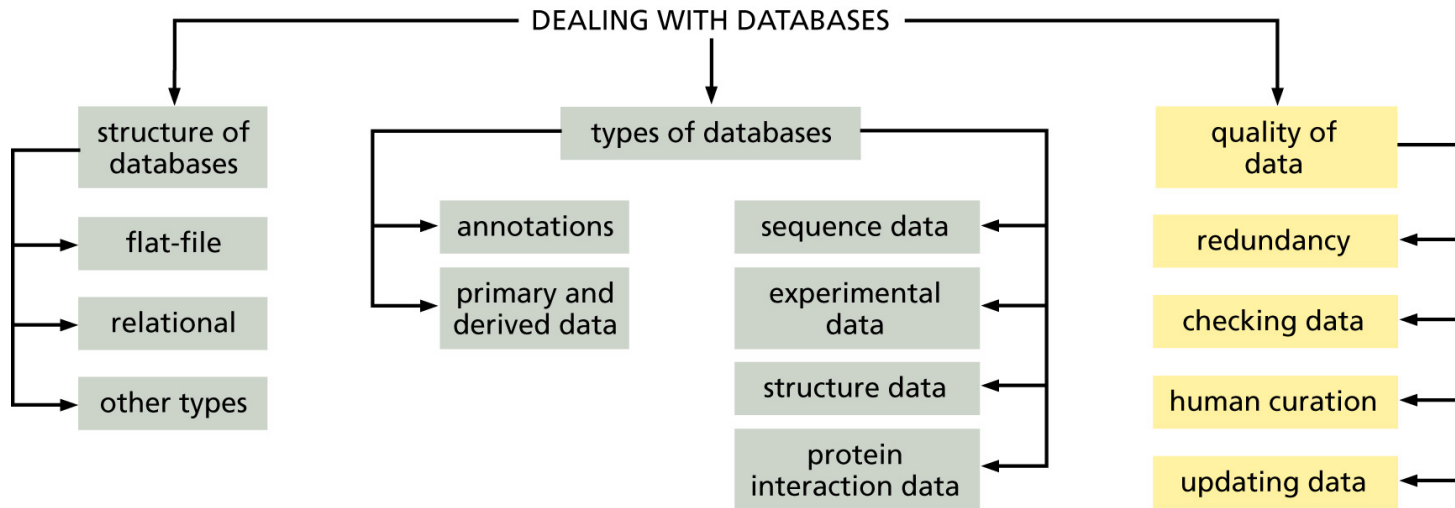
Non reduntant databases: they include all the experimental data (from different labs) in one entry.

Checking data: a DNA seqeunce 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

Useful for:

-comparing an unknown sequence to all the sequences contained in a database;

- prediction of a protein structure

- construction of phylogenetic trees
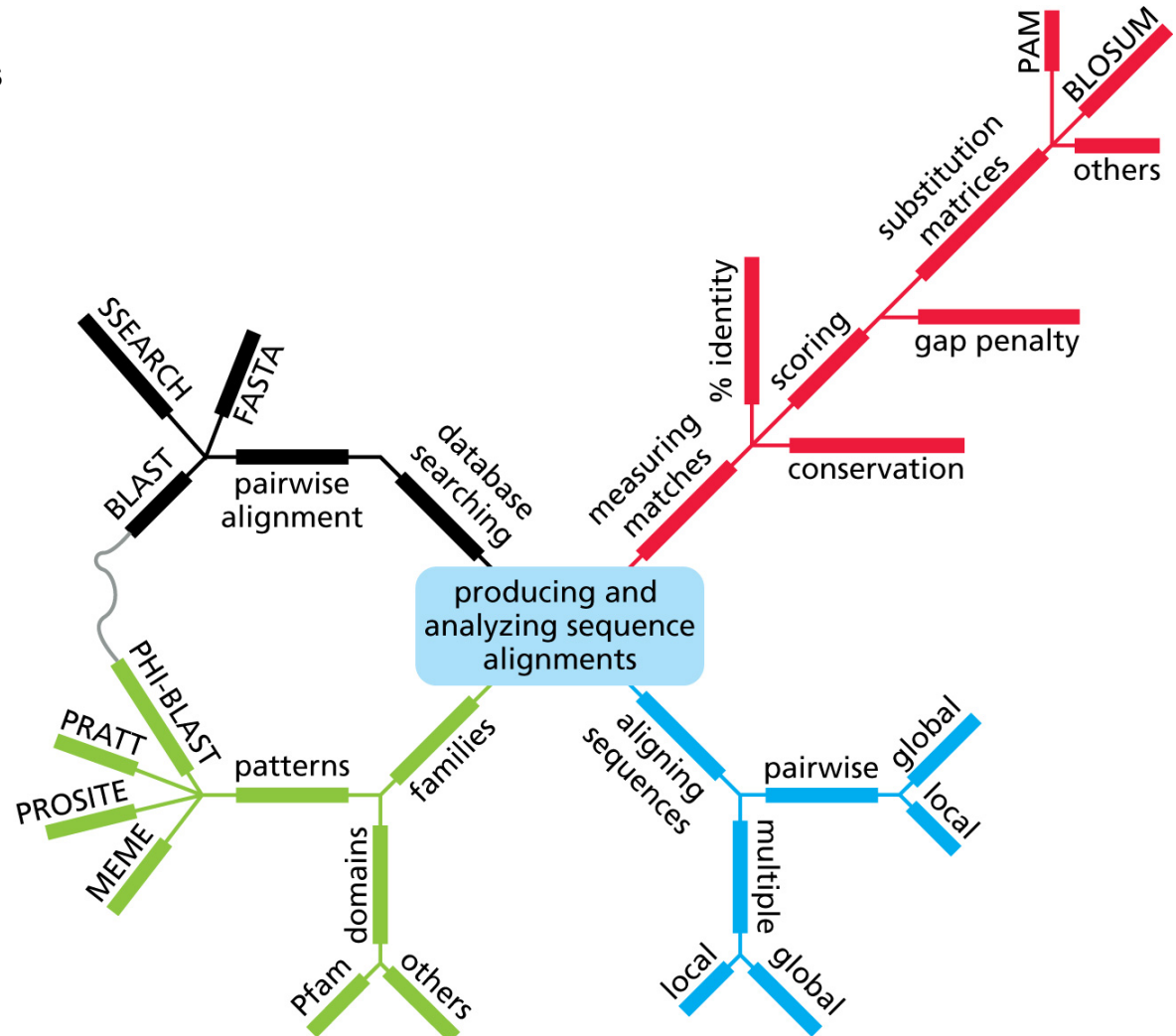
Sequence alignments

Alignment is the task of locating equivalent regions of two or more sequences to maximize their similarity.

T H I S S E Q U E N C E

T H A T S E Q U E N C E

The differences in length between two or more sequences can be compensated by the introduction of **GAPS**.

T H I S I S A - S E Q U E N C E

T H - - - - A T S E Q U E N C E

Gap penalty: each time a gap is introduced, the penalty is subtracted from the score, decreasing the overall score of the alignment.

(A)

```
Bovine PI-3Kinase p110a        LNWENPDIMSELLFQNNEIIFKNGDDLRQDMLTLQIIRIMENIWQNQGLDLRMLPYGCLSIGDCVGLIEVVRNSHTIMQIQCKGGLKGAL
cAMP-dependent protein kinase  --WENPAQNTAHLDQFERIKTLGTGSFGRVMLVKHMETGNHYAMKILDKQKVVKLKQIEHTLNEKRILQAVNFPFLVKLEFSFKDNSNLY

Bovine PI-3Kinase p110a        QFNSHTLHQWLKDKNKGEIYDAAIDLFTRSCAGYCVATFILGIGDRHNSNIMVKDDGQLFHIDFGHFLDHKKKFGYKRERVPFVLTQDF
cAMP-dependent protein kinase  MVMEYVPGGEMFSHLRRIGRFSEPHARFYAAQIVLTFEYLHSLDLIYRDLKPENLLIDQQGYIQVTDFGFAKRVKGRTWXLCGTPEYLAP

Bovine PI-3Kinase p110a        LIVISKGAQECTKTREFERFQEMCYKAYLAIRQHANLFINLFSMMLGSGMPELQSFDDIAYIRKTLALDKTEQEALEYFMKQMNDAHHGG
cAMP-dependent protein kinase  EIILSKGYNKAVDWWALGVLIYEMAAGYPPFFADQPIQIYEKIVSGKVRFPSHFSSDLKDLLRNLLQVDLTKRFGNLKNGVNDIKNHKWF

Bovine PI-3Kinase p110a        WTTKMDWIFHTIKQHALN----------------------------------
cAMP-dependent protein kinase  ATTDWIAIYQRKVEAPFIPKFKGPGDTSNFDDYEEEEIRVXINEKCGKEFSEF
```

(B)

```
Bovine PI-3Kinase p110a        LNWENPDIMSELLFQNNEIIFKNGDDLRQDMLTLQIIRIMENIWQNQGLDLRMLPYGCLSIGDCVGLIEVVRNSHTIMQIQCKGGLKGAL
cAMP-dependent protein kinase  ?-WENPAQNTAHLDQFERIKTLGTGSFGRVMLVKHM--ETGNHYAMKILDKQKV-VKLKQIEHTLNEKRILQAVNFPFLVKLEFSFKDN-

Bovine PI-3Kinase p110a        QFNSHTLHQWLKDKNKGEIYDAAIDLFTRSCAGYCVATFILGIGDRHNSNIMVKD-DGQLFHIDFGHFLDHKKKKFGYKRERVPFVL--T
cAMP-dependent protein kinase  -SNLYMVMEYVPGGEMFSHLRR-IGRFSEPHARFYAAQIVLTFEYLHSLDLIYRDLKPENLLIDQQGYIQVTDFGFAKRVKGRTWXLCGT

Bovine PI-3Kinase p110a        QDFL---IVISKGAQECTKTREFERF-QEMC--YKAYLAIRQHANLFINLFSMMLGSGMPELQSFDDIAYIRKTLALDKTEQEALEYFMK
cAMP-dependent protein kinase  PEYLAPEIILSKGYNKAVDWWALGVLIYEMAAGYPPFFA-DQPIQIYEKIVSGKVRF--PSHFSSDLKDLLRNLLQVDLTKR--FGNLKN

Bovine PI-3Kinase p110a        QMNDAHHGGWTTKMDWI----------------------FHTIKQHAL----N----------
cAMP-dependent protein kinase  GVNDIKNHKWFATTDWIAIYQRKVEAPFIPKFKGPGDTSNFDDYEEEEIRVXINEKCGKEFSEF
```

A) An alignment where the gap penalty has been set very high.

B) An alignment with a very long gap penalty. Many more gaps have been introduced.

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 ⇒ common ancestor ⇒ common structure or function?
Not always……

PRODUCING AND ANALYZING
SEQUENCE ALIGNMENTS

looking for
similarity

assessing
similarity

scoring
measures

gaps in
alignment
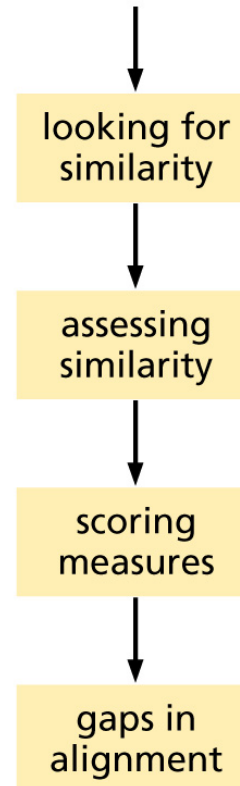
Sequence alignments

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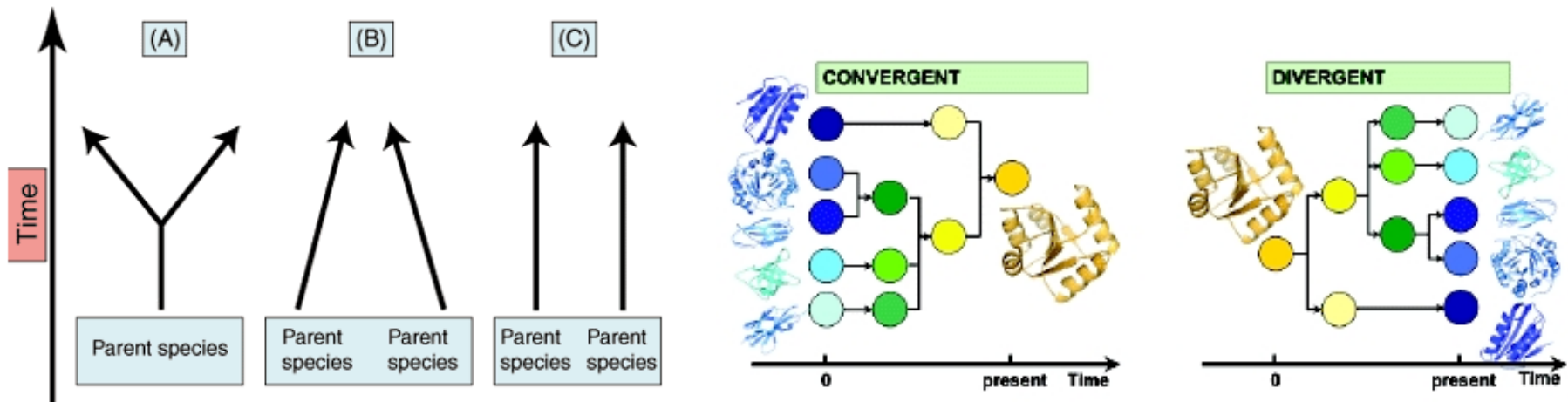
It is easier to compare to detect homology when comparing protein sequence than when comparing nucleic acid sequences.

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

looking for
similarity

assessing
similarity

scoring
measures

gaps in
alignment
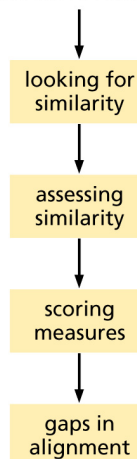
Scoring alignments: identity percentage and similarity percentage

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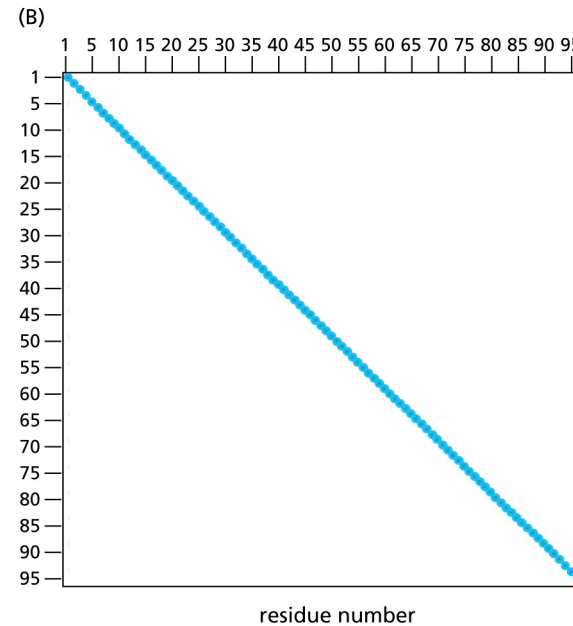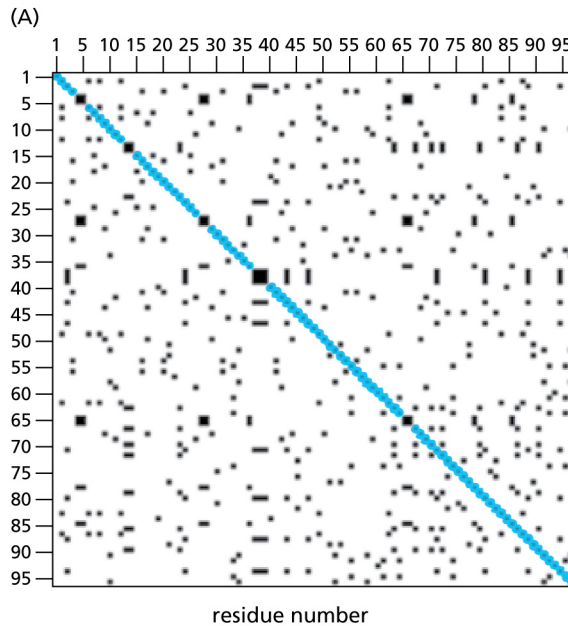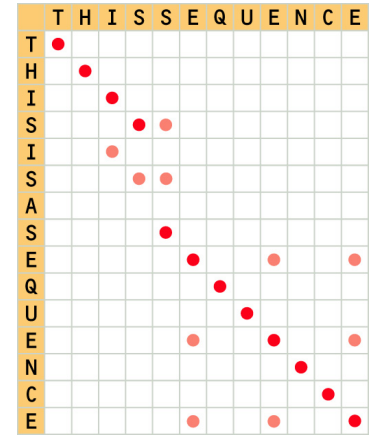
Two views of dot-plot representations of an SH2 sequence compared to itself. A) Unfiltered dot-plot. The identity is shown by the unbroken diagonale. There is some background noise. B) Dot-plot of the same sequence comparison with a window of 10 residues and a minimum identity score within the window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

Scoring alignments: identity percentage and similarity percentage

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

T H I S I S A - S E Q U E N C E

T H - - - - - A T S E Q U E N C E

T H I S I S A - S E Q U E N C E
T H A T - - - - S E Q U E N C E

```
gi|66361410|pdb|1ZBM|A            --------------------------MGHHHHHHSHKIRVAHTPDADD  22
gi|154175534|ref|YP_001409022.    ------------------------------MKNIKHIDVAHSPDADD  17
gi|6647837|sp|O28098.1|SUCD2_A    --------------------------MAIIVDERTKVVVQGITGYQGK  22
gi|1711576|sp|P53598.1|SUCA_YE    MLRSTVSKASLKICRHFHRESIPYDKTIKNLLLPKDTKVIFQGFTGKQGT  50
                                                  .  :: . :  .  .

gi|66361410|pdb|1ZBM|A            AFXFYAXTHGKVDT-WLEIEHVIEDIETLNRKAFNAEYEVTAISAHAYAL  71
gi|154175534|ref|YP_001409022.    IFMYMAIKFGWVGSKNLSFTNTALDIQTLNEEALKSTYTATAISFALYPL  67
gi|6647837|sp|O28098.1|SUCD2_A    FHTERMLNYGTKIVAGVTPGKGGTEVLGVPVYDSVKEAVREADANASVIF  72
gi|1711576|sp|P53598.1|SUCA_YE    FHASISQEYGTNVVGGTNPKKAGQTHLGQPVFASVKDAIKETGATASAIF 100
                                     .     .*           :                :  :     :

gi|66361410|pdb|1ZBM|A            LDDKYRILSAGASVGDGYGPVVVAKSEISLD-GKRIAVPGRYTTANLLLK 120
gi|154175534|ref|YP_001409022.    ISDDYALLRCAVSFGEGYGPKLIKKRGVNLKRNFKVALSGAHTTNALLFR 117
gi|6647837|sp|O28098.1|SUCD2_A    VPAPFAADAVMEAADAGIKVIVCITEGIPVHDELKMYWRVKEAGAT-LIG 121
gi|1711576|sp|P53598.1|SUCA_YE    VPPPIAAAAIKESIEAEIPLAVCITEGIPQHDMLYIAEMLQTQDKTRLVG 150
                                    :          :        :  . :  .        :          *.

gi|66361410|pdb|1ZBM|A            LAVE-DFEPVEXPFDRIIQAVLDEEVDAGLLIHEGQITYADYGLKCVLDL 169
gi|154175534|ref|YP_001409022.    AAYP-EARIVYKNFLEIENAVLSGEVDAGVLIHESILGFSS-ELEVEREI 165
gi|6647837|sp|O28098.1|SUCD2_A    PNCPGIISPG-KTHLGIMPVQIFKPGNVGIVSRSGTLTYQIAYNLTKLGL 170
gi|1711576|sp|P53598.1|SUCA_YE    PNCPGIINPATKVRIGIQPPKIFQAGKIGIISRSGTLTYEAVQQTTKTDL 200
                                     *    :    . *::  :.. : :              :

gi|66361410|pdb|1ZBM|A            WDWWSEQV--KLPLPLGLNAIRRDLSVEVQEEFLRAXRESIAFAIEN-PD 216
gi|154175534|ref|YP_001409022.    WDVWCELAGENLPLPLGGMALRRSLPLTDAIECERVLTKAVAIATAHKPF 215
gi|6647837|sp|O28098.1|SUCD2_A    GQSTVVGIGGDRIIGTDFVEVLRLFEDDKETKAVVLVGEIGGRDEEVAAE 220
gi|1711576|sp|P53598.1|SUCA_YE    GQSLVIGMGGDAFPGTDFIDALKLFLEDETTEGIIMLGEIGGKAEIEAAQ 250
                                     :          .        :: :           :      :   .

gi|66361410|pdb|1ZBM|A            EAIEYAX---------KYSRGLDRERAKRFAXXYVNDYTYNXPESVDAAL 257
gi|154175534|ref|YP_001409022.    LSHMLME---------RNLIRIDKEKLKIYLNLYANKDSISMNETQLKAL 256
gi|6647837|sp|O28098.1|SUCD2_A    FIREMS------KPVVGYVAGLTAPPGK--RMGHAGAIIEGGVGTAESKI 262
gi|1711576|sp|P53598.1|SUCA_YE    FLKEYNFSRSKPMPVASFIAGTVAGQMKGVRMGHSGAIVEGSGTDAESKK 300
                                                          *       :  .          .

gi|66361410|pdb|1ZBM|A            KKLYEX----------AEAKGLIKMPKLDILRL--  280
gi|154175534|ref|YP_001409022.    NRLFEIGYDQGFYPQPIDAHDYLIPTEYNDARFS- 290
gi|6647837|sp|O28098.1|SUCD2_A    KALEAAG------ARVGKTPMEVAELVAEIL---- 287
gi|1711576|sp|P53598.1|SUCA_YE    QALRDVG------VAVVESPGYLGQALLDQFAKFK 329
                                   : *         .:   :    :
```

Scoring alignments: substitution matrices

(A)

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

(B)

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

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
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.	blastp/blastn
Uses Smith–Waterman algorithm. Can search protein to protein or DNA to DNA. Can be more sensitive than fasta with protein sequences.	
DNA compared to protein database. DNA translated into all three frames. fasty slower than fastx but better. Used to see if DNA encodes a protein.	blastx
Protein compared to DNA database. Mainly used to identify EST sequences. This is preferred over fastx as protein comparison is more sensitive than DNA.	tblastn (tblastx compares translated DNA to translated DNA database)
Mixed peptide sequence (such as obtained by Edman degradation) compared to protein database.	
Mixed peptide sequence compared to DNA database.	

(A)

sp|P32871|P11A_BOVIN	PHOSPHATIDYLINOSITOL 3-KINASE CATALYTI...	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|RIR1_MYCGE	RIBONUCLEOSIDE-DIPHOSPHATE REDUCTASE A...	34	0.59

(B)

Distribution of 2 Blast Hits on the Query Sequence

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

Color Key for Alignment Scores

<40 40-50 50-80 80-200 >=200

tmpseq_1   0   50   100   150   200   250   300

dbj|BAB10275.1|	(AB008266) phosphatidylinositol 4-kinase [A...	111	3e-25
dbj|BAB11344.1|	(AB011477) AtRAD3 [Arabidopsis thaliana]	38	0.008
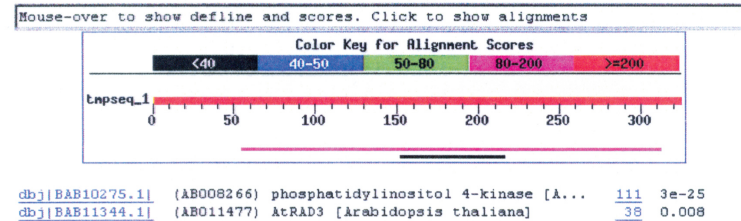
(C)

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

Mouse-over boxes to display more information

1   50   100   150   200   250   260

PI3Kc

PI3_PI4_kinase

Sequences producing significant alignments:

		Score (bits)	E value
gnl|Smart|PI3Kc	Phosphoinositide 3-kinase, catalytic domain; Phosphoinositide ...	301	3e-83
gnl|Pfam|pfam00454	PI3_PI4_kinase, Phosphatidylinositol 3- and 4-kinases	263	9e-72

gnl\|Smart\|PI3Kc, Phosphoinositide 3-kinase, catalytic domain; Phosphoinositide 3-kinase isoforms participate in a variety of processes, including cell motility, the Ras pathway, vesicle trafficking and secretion, and apoptosis. These homologues may be either lipid kinases and/or protein kinases: the former phosphorylate the 3-position in the inositol ring of inositol phospholipids. The ataxia telangiectasia-mutated gene produced, the targets of rapamycin (TOR) and the DNA-dependent kinase have not been found to possess lipid kinase activity. Some of this family possess PI-4 kinase activities.

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 IIFKNGDDLRQDMLTLQIIRIMENIWQNQGLDLRMLPYGCLSIGDCVGLIEVVRNSHTIM 78
Sbjct: 2 IIFKHGDDLRQDMLILQILRIMESIWETESLDLCLLPYGCISTGDKIGMIEIVKDATTIA 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.
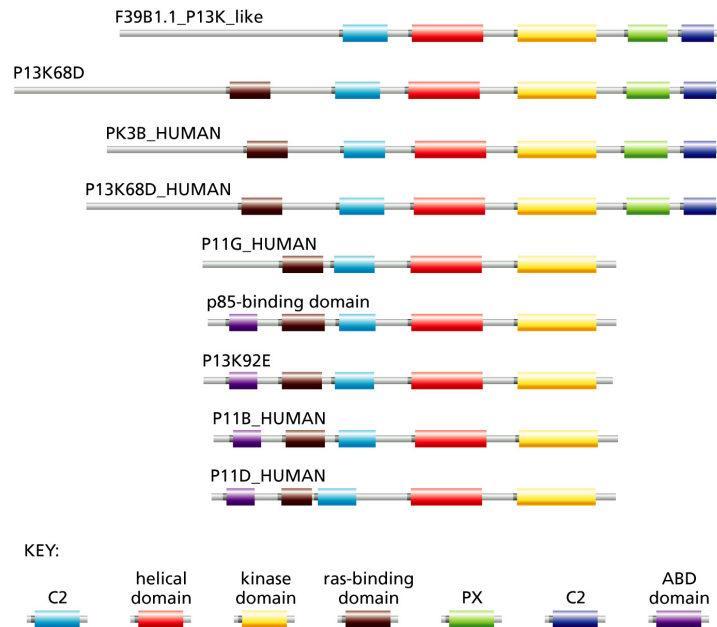
PI3-kinase is a multidomain protein. Output from Pfam.

Multiple alignments

They can be constructued by different techniques.

(A)
structural/functional alignment from BAliBase

```
1csy   SHEKMPWFHGKISREESEQIVLIGSKTNGKFLIRARD--NNGSYALCLLHEGKVLHYRIDKDKTGKLSIPEGK-KFDTLWQLVEHYSYKA------DGLLRVL-TVPCQK
1gri   EMKPHPWFFGKIPRAKAEEML-SKQRHDGAFLIRESES-APGDFSLSVKFGNDVQHFKVLRDGAGKYFL-WVV-KFNSLNELVDYHRST-VSRNQQIFLRDIEQVPQQ-
1aya   ---MRRWFHPNITGVEAENLLLTRG-VDGSFLARPSKS-NPGDFTLSVRRNGAVTHIKIQN--TGDYYDLYGGEKFATLAELVQYYMEHHGQLKEKNGDVIEL-KYPLN-
2pna   -LQDAEWYWGDISREEVNEKLRDT--ADGTFLVRDASTKMHGDYTLTLRKGGNNKLIKIFH-RDGKYGFSDPL-TFNSVVELINHYRNES-LAQYNPKLDVKL-LYPVS-
1bfi   HHDEKTWNVGSSNRNKAENLLRGK--RDGTFLVRESS--KQGCYACSVVVDGEVKHCVINKTATG-YGFAEPYNLYSSLKELVLHYQHTS-LVQHNDSLNVTL-AYPVYA
```

(B)
DIALIGN multiple sequence alignment

```
1csy   SHEKMPWFHGKISREESEQIVLIGSKT-NGKFLIRAR-DN--NGSYALCLLHEGKVLHYRIDKDKTGKLSIPEGKK-FDTLWQLVEHYSYKA-------DGLLRVLT-VPCQK
1gri   EMKPHPWFFGKIPRAKAEEML--SKQRHDGAFLIRESESA--PGDFSLSVKFGNDVQHFKVLRDGAGKYFLWVV--K-FNSLNELVDYHRST--SVSRNQQIFLRDIEQVPQQ-
1aya   M---RRWFHPNITGVEAENLLLTRGV--DGSFLARPSKSN--PGDFTLSVRRNGAVTHIKIQNTGDYYDLYG-GEK-FATLAELVQYYMEHHGQLKEKNGDV-IELK-YPLN-
2pna   LQDAE-WYWGDISREEVNEKL--RDTA-DGTFLVRDA-STKMHGDYTLTLRKGGNNKLIKIFHRDGKYGFSD-PLT-FNSVVELINHYRNE--SLAQYNPKLDVKLL-YPVS-
1bfi   HHDEKTWNVGSSNRNKAENLL--RGKR-DGTFLVRES-SK--QGCYACSVVVDGEVKHCVINKTATGYGFAE-PY-NLYSSLKELVLHYQHT--SLVQHNDSLNVTLA-YPVYA
```

(C)
ClustalW multiple sequence alignment

```
1csy   SHEKMPWFHGKISREESEQIVLIGSKTNGKFLIRARDN--NGSYALCLLHEGKVLHYRIDKDKTGKLSIPEGKKFD-TLWQLVEHYSYK------ADGLLRVLTVPCQK
1gri   EMKPHPWFFGKIPRAKAEE-MLSKQRHDGAFLIRESES-APGDFSLSVKFGNDVQHFKVLRDGAGKY-FLWVVKFN-SLNELVDYHRSTS-VSRNQQIFLRDIEQVPQQ
1aya   ---MRRWFHPNITGVEAEN-LLLTRGVDGSFLARPSKS-NPGDFTLSVRRNGAVTHIKIQNT-GDYYDLYGGEKFA-TLAELVQYYMEHHGQLKEKNGDVIELKYPLN-
2pna   -LQDAEWYWGDISREEVN--EKLRDTADGTFLVRDASTKMHGDYTLTLRKGGNNKLIKIFHR-DGKYGFSDPLTFN-SVVELINHYRNES-LAQYNPKLDVKLLYPVS-
1bfi   HHDEKTWNVGSSNRNKAE--NLLRGKRDGTFLVRESSK--QGCYACSVVVDGEVKHCVINKT-ATGYGFAEPYNLYSSLKELVLHYQHTS-LVQHNDSLNVTLAYPVYA
```

(D)
divide-and-conquer multiple sequence alignment

```
1csy   SHEKMPWFHGKISREESEQIVLIGSKTNGKFLIRA-RDNN-GSYALCLLHEGKVLHYRIDKDKTGKLSIPEGKK-FDTLWQLVEHY-SY----KADGLLRV-L-TVPCQK
1gri   EMKPHPWFFGKIPRAKAEEMLS-KQRHDGAFLIRE-SESAPGDFSLSVKFGNDVQHFKVLRDGAGK-YFLWVVK-FNSLNELVDYH-RSTSVSRNQQIFLRDIEQVPQQ-
1aya   ---MRRWFHPNITGVEAENLLL-TRGVDGSFLARP-SKSNPGDFTLSVRRNGAVTHIKIQNTGDYY-DLYGGEK-FATLAELVQYYMEHHGQLKEKNGDVIEL-KYPLN-
2pna   -LQDAEWYWGDISREEVNEKL--RDTADGTFLVRDASTKMHGDYTLTLRKGGNNKLIKIFHRDGKY-GFSDPLT-FNSVVELINHY-RNESLAQYNPKLDVKL-LYPVS-
1bfi   HHDEKTWNVGSSNRNKAENLL--RGKRDGTFLVRE-SSKQ-GCYACSVVVDGEVKHCVINKTATG-GFAEPYNLYSSLKELVLHY-QHTSLVQHNDSLNVTL-AYPVYA
```

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

OBTAINING SECONDARY STRUCTURE FROM SEQUENCE
→ different ways to predict → nearest neighbor / statistical methods / machine learning → neural nets / HMMs → mixture of the above
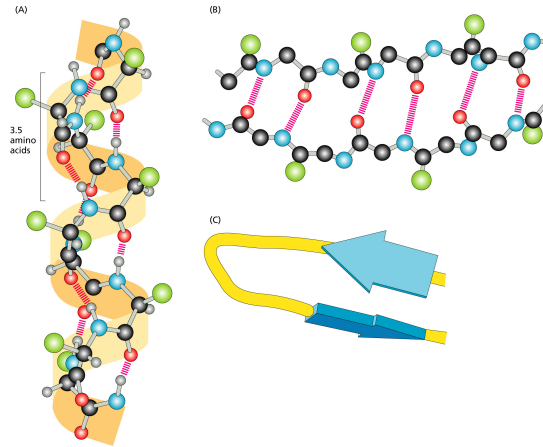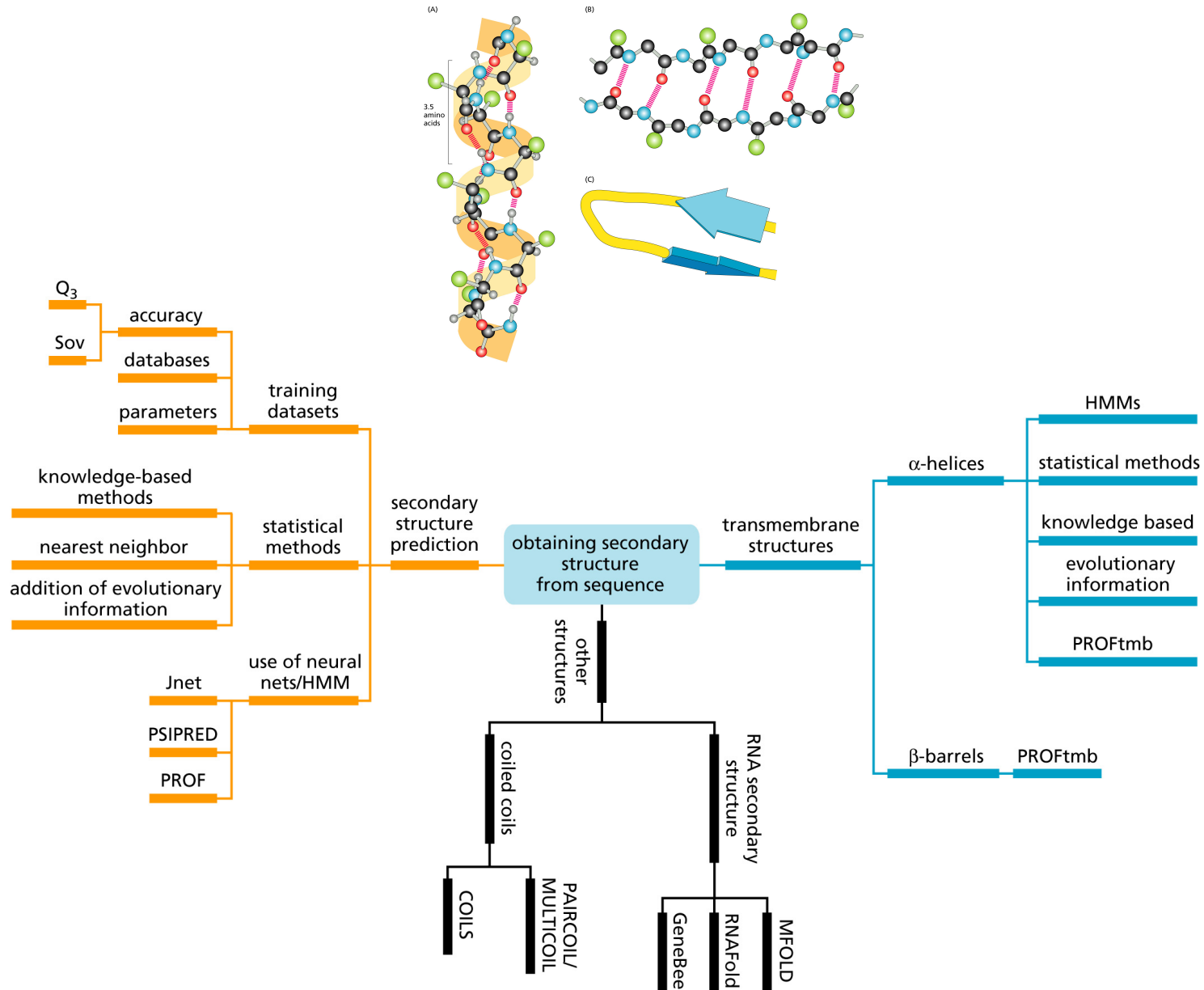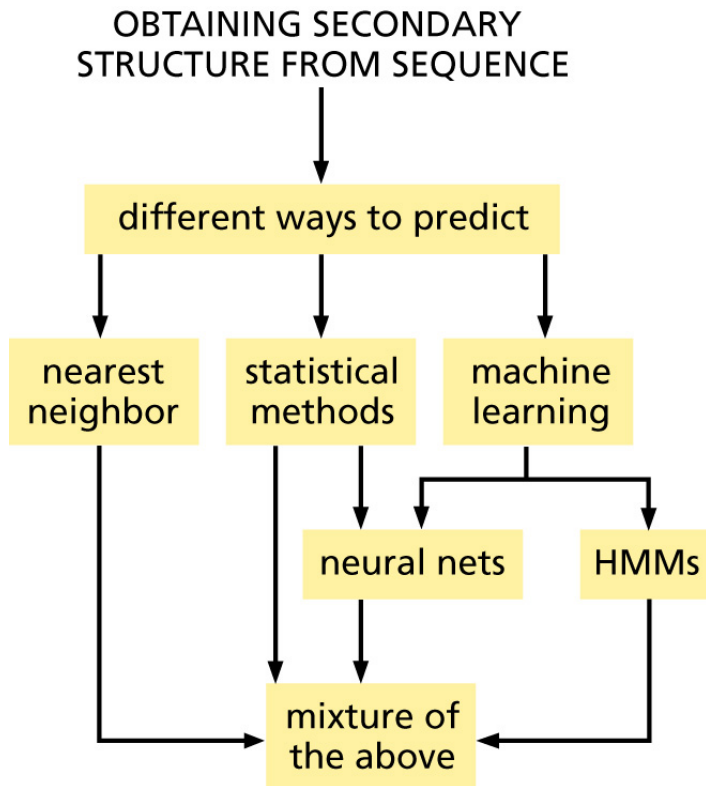
Statistical methods are based on rules that give the probability that a residue will form part of a particular secondary structure.

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 alghoritms to aquire 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
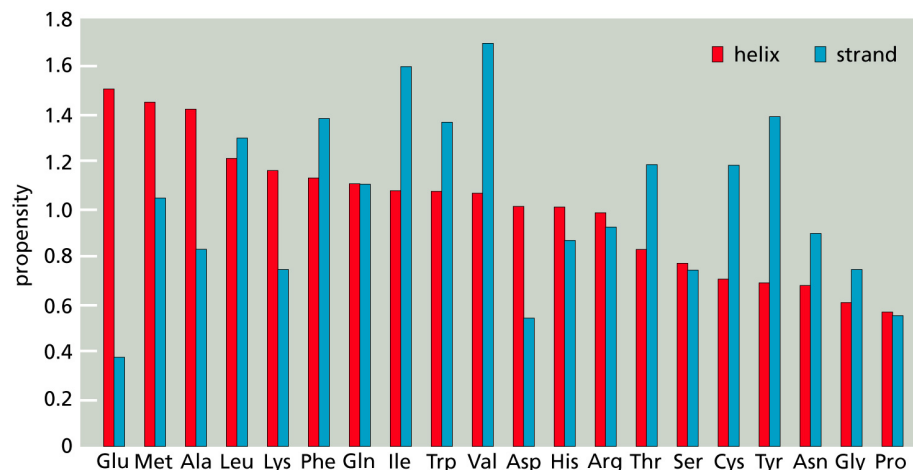
Chou-Fasman is one of most commonly used algorithms

- 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**
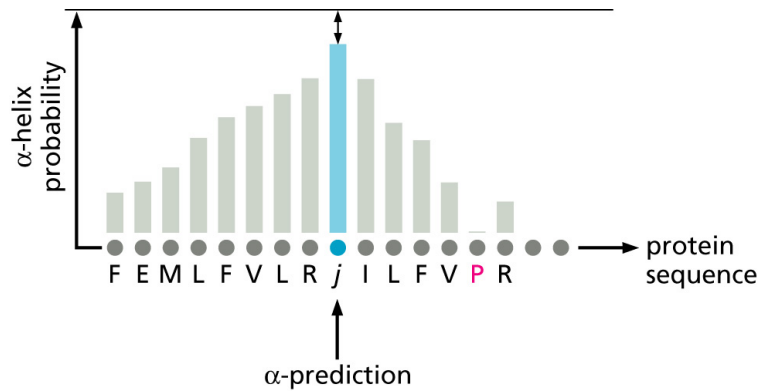 4. **f(i+3) = probability of being in position 4**

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.
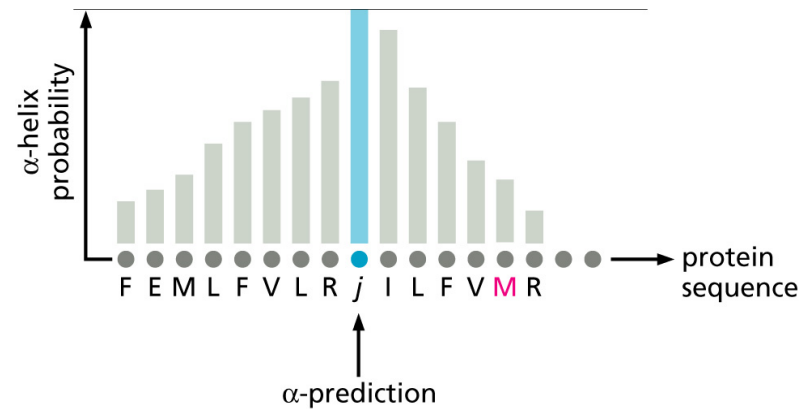
- helix: 4 out of 6 amino acids with P(a) >100
  - extends the nucleus in each direction until reach four amino acids in a row with P(a) <100
  - for each of these regions, add up all the P(a) and all the P(b) values.
  - If the total P(a) is larger than the total of P(b) and the run is more than 5 amino 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(turn) for amino acids i to i+3 is > 100
    - the sum of the P(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.

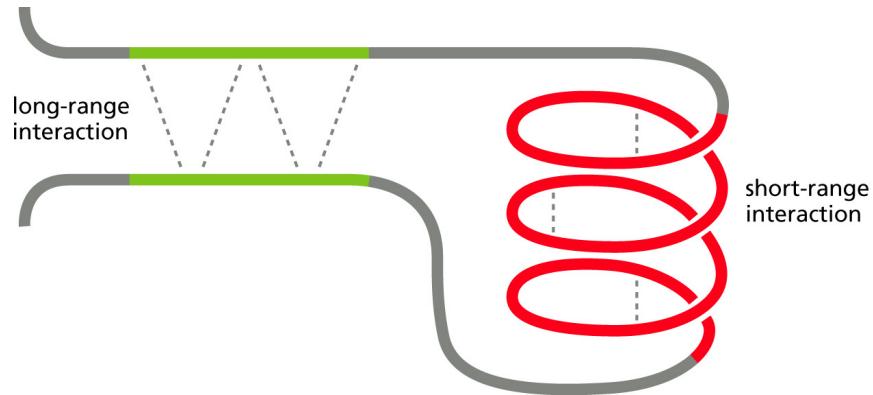The effect of an helix breaker (Pro) at position j+5. The proline diminishes the overall additive propensity of residue j to form helix

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

Statistical methods improvements: GOR I to V

1B8C

```
1B8C    AFAGVLNDADIAAALEACKAADSFNHKAFFAKVGLTSKSADDVKKAFAIIAQDKSGFIEEDELKLFLQNFKADARALTDGETKTFLKAGDSDGDGKIGVDDWTALVKA
GOR I   HHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHCCCHHHHHHHHHHHHHHHHHHHHHHHHHHHHEHCTCTTTEEEEEEHHHHHHHC
GOR IV  CCCCCCHHHHHHHHHHHHCCCCCHHHHEEECCCCCCHHHHHHHHHHHHCCCCCHHHHHHHHHHHHHHHHHHHCCCCCEEEEEECCCCCCCCCEEECCCEEEEEEC
GOR V   CCCCCCCHHHHHHHHHCCCCCCCCHHHHHHHHHHHHHHHHHHHCCCCCCHHHHHHHHHHHHHCCCCCCCCCHHHHHHHHHHCCCCCCCCCHHHHHCCCCC
X-RAY   CCBTTBTHHHHHHHHHHTTTTTCCHHHHHHHTTCTTSCHHHHHHHHHHHHTSTTCEECHHHHTTTGGGTTTTCCCCHHHHHHHHHHHCSSCSSSEHHHHHHHHTT
```

1BKB

```
1BKB    KWVXSTKYVEAGELKEGSYVVIDGEPCRVVEIEKSKTGKHGSAKARIVAVGVFDGGKRTLSLPVDAQVEVPIIEKFTAQILSVSGDVIQLXDXRDYKTIEVPXKYVEEEAKGRLAPGAEVEVWQILDRYKIIRVKG
GOR I   HHHHEEEHHHHHHHHHHHEEEEECCHHHHHHHHHHHHHHHHHHHEEEEEEEETTTTEEEEEEEEHHHHHEHHHHHHHHHEEEECEEEEEEEEHHTTTEEEEEHHHHHHHHHHHHHHHCHHHHHHHHHHHTEEEEEET
GOR IV  CCEEEEEEEECCCCCCCEEEEECCCCCEEEEECCCCCCCCCCHHHHEEEEEEECCCCCEEECCCCCCCCCHHHHHCHHHHHCEECEEEEEEEECCEEEEEECHHHHHHHHHHCCCCCCHHHHHHCCCCEEEEEC
GOR V   CCCCCCCCCCCCCCCCCEEEEECCCCEEEEEEECCCCCCCCCEEEEEEEEEEECCCCCEEEECCCCCCCHHHHHHHHHEEEECCCCCEEEECCCCHHHHCCCCCHHHHHHHHHHCCCCEEEECCCCCCCCCCCC
X-RAY   CCCCCCEEEGGGTTTTCEEEEETTEEEEECEEEEECCSTTSCCEEEEEEEEETTTCEEEEEEETTSEEECCCEEEEEEEECEECSSEEEEEETTTCCEEEEEGGGBTHHHHTTTTTTCEEEEEEEETTEEEEECEECC
```

1CJW

```
1CJW    HTLPANEFRCLTPEDAAGVFEIEREAFISVSGNCPLNLDEVQHFLTLCPELSLGWFVEGRLVAFIIGSLWDEERLTQESLALHRPRGHSAHLHALAVHRSFRQQGKGSVLLWRYLHHVGAQPAVRRAVLMCEDALV
GOR I   ECCCTHHHEEECHHHHHHHHHHHHHHHHHETTTTCCCHHHHHHHEEEETHHHHHHHHHHHHEEEECCCCHHHHHHHHHHHHHTTTHHHHHHHHHHHHHHTTTTCEEEEEHEEECCTCEHHHHHHHHHHHHHHH
GOR IV  CCCCCCCCCCCCCCCCCCHHHHHHHEEEEECCCCCCCCCCCCHHHHCCCCCCCHHHCCEEEEEEECCCHHHHHHHHHHHHCCCCCHHHHHHHHHHHHHHCCCCCCEEEEEEEEECCCCCHHHHHHHHHCCCCCC
GOR V   CCCCCCCCCCCCCCCCHHHHHHHHHHHCCCCCCCCCCCCHHHHHHHCCCCCEEEEEECCCCEEEEEEECCCCCCCCCCCCCCCCCCCCCCCCCEEEEEEECCCCCCCCCCHHHHHHHHHHHHHHHEEEEECCCCHHH
X-RAY   CCCCSSEEECCCGGGHHHHHHHHHHHTHHHHSCCCCHHHHHHHHCGGGEEEEEEETTECEEEEEEEEECCCCCGGGGGCCCTTCEEEEECEEEECTTCCCCCHHHHHHHHHHHHHHTTTTCCEEEEEECGGGH
```

```
1CJW    PFYQRFGFHPAGPCAIVVGSLTFTEMHCSL
GOR I   HHEEETTTCTTCTEEEEEEECHHHHHHHH
GOR IV  CCCCCCCCCCCCCEEEECCEEEEEECCEEC
GOR V   HHHHHCCCCCCCCCCCCCCCCCCCCCCCCC
X-RAY   HHHHTTTEEECCCCCCCCCCCCCEEEEEEC
```

1CT5

```
1CT5    STGITYDEDRKTQLIAQYESVREVVNAEAKNVHVNENASKILLLVVSKLKPASDIQILYDHGVREFGENYVQELIEKAKLLPDDIKWHFIGGLQTNKCKDLAKVPNLYSVETIDSLKKAKKLNESRAKFQPDCNPI
GOR I   EEEEEEEHHHHHHEEEEHEHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHCHHHHEEEEEECCEEHHCHHHHHHHHHHHHHHHHHHHHEEEETTTCTTHEHHHEEEEEEEEEEHHHHHHHHHHHHHHHHHHHHTEETTTCTE
GOR IV  CCCCCCCHHHHHHHHHHHHHHHHHHHHCCCEEECCCHHHHHHHHHCCCCCCHHHHHCCCCCCHHHHHHHHHHHCCCCCCEEEEECCCCCCCCCCCCCCCCCCCHHHHHHHHHHHHHHHHHHHCCCCCCCE
GOR V   CCCCCCCCCCCCCCCCCHHHHHHHHHHHHHHHHCCCCCCEEEEEECCCCCHHHHHHHHHCCCCCCCHHHHHHHHHCCCCEEEEEECCCCCCHHHHHHHHCEEEEECHHHHHHHHHHHHHHHHHCCCCCCE
X-RAY   CCCCCCHHHHHHHHHHHHHHHHHHHHHHTCCCCCCCCCEEEEECCTTSHHHHHHHHHHHTCCEEEECHHHHHHHHHHSCTTCEEEECSCCCGGGHHHHHCTTEEEEEECSHHHHHHHHHHHHHHHCTTSCCE
```

```
1CT5    LCNVQINTSHEDQKSGLNNEAEIFEVIDFFLSEECKYIKLNGLMTIGSWNVSHEDSKENRDFATLVEWKKKIDAKFGTSLKLSMGMSADFREAIRQGTAEVRIGTDIFGARPPKNEARII
GOR I   EEEEEEEEEHHTTTCCCCHHHHHHHHHHHHHHHHHEEEEEETCCCCCCTHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHHCHEEEEEEEEEEEEECCCCCHHHHEE
GOR IV  ECECCCCCCCCCCCCCCCHHHHHHHHHHCCCCCEEEEECEEEEEECCCEECCCCCCCCCHHHHHHHHHHHHHCCCHHHHHHHHHHHHHHHHHCCEEEEECCCCCCCCCCCCCEEEC
GOR V   EEEEEEECCCCCCCCCCCCCHHHHHHHHHHHHCCHHHHHEEECCCCCCCCHHHHHHHHHHHHHHHHHCCCCHHHHHCCCCHHHHHHHCCEEEEEEEEECCCCCCCCCCCCCC
X-RAY   EEEEEBCCSSSCCSSSBCHHHHHHHHHHHHHSTTCCSEEEEEEECCCCCCCCCCCCCCHHHHHHHHHHHHHHHHCCCCEEECCCTTTHHHHHHTTCSEEEESHHHCCCCCCCCCCCCCCC
```

Nearest neighbor methods

The formation of secondary structure in proteins does not only depend on local interactions (beta-sheets are made up of beta-strands that are separated from some distance in the poypeptide chain).

long-range interaction

short-range interaction

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	-1	4	-2	2	0	-4	-2	2
0	-3	-1	-2	0	-2	4	-3	-3	0	-2	-2	-4	-3	3	1	-4	-4	-4	-3
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	-3	-2	-3	-3	-2	-3	1	-2	-3	-2	1	-3	0	1	-4	-2	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

window of 15 rows

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

15 x 20 scaled inputs to 1st network

1st network
315 input units
75 hidden layers
3 outputs

H, E, L

2nd network
60 input units
60 hidden layers
3 outputs

H, E, L

N-terminal...T H I S I S A H I D D E N M E S S A G E ...C-terminal

input layer: 0 0 0 1 0 0 0 0

hidden layer

output layer: α   β   coil

firing result: 0.7   0.1   0.2

prediction result: α

Conf:

Pred:

Pred: CCCCCCCCCCCCHHHHHHHHHCCCEEEEECCCCCCCCHHHHHHHHHCCCCEEEEECCCEEEEEEECCCCCCCCCEECCCH
AA: HTLPANEFRCLTPEDAAGVFEIEREAFISVSGNCPLNLDEVQHFLTLCPELSLGWFVEGRLVAFIIGSLWDEERLTQESL
      10   20   30   40   50   60   70   80
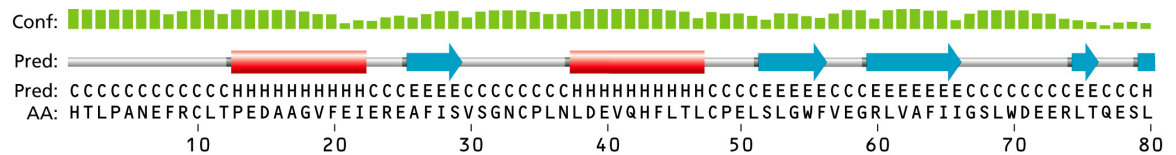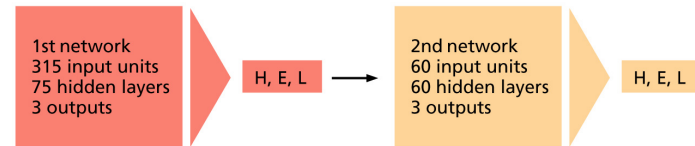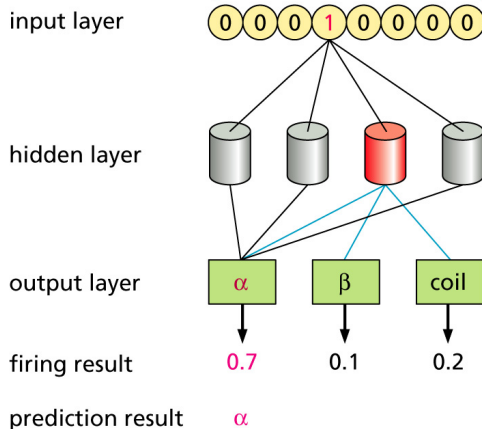
Secondary structure prediction methods

(A)

1B8C

X-ray:

GOR IV:

GOR V:

PredS:

PredM:

Zpred:

PROF:

NNSSP:

PHD:

PSIPRED:

Jnet:
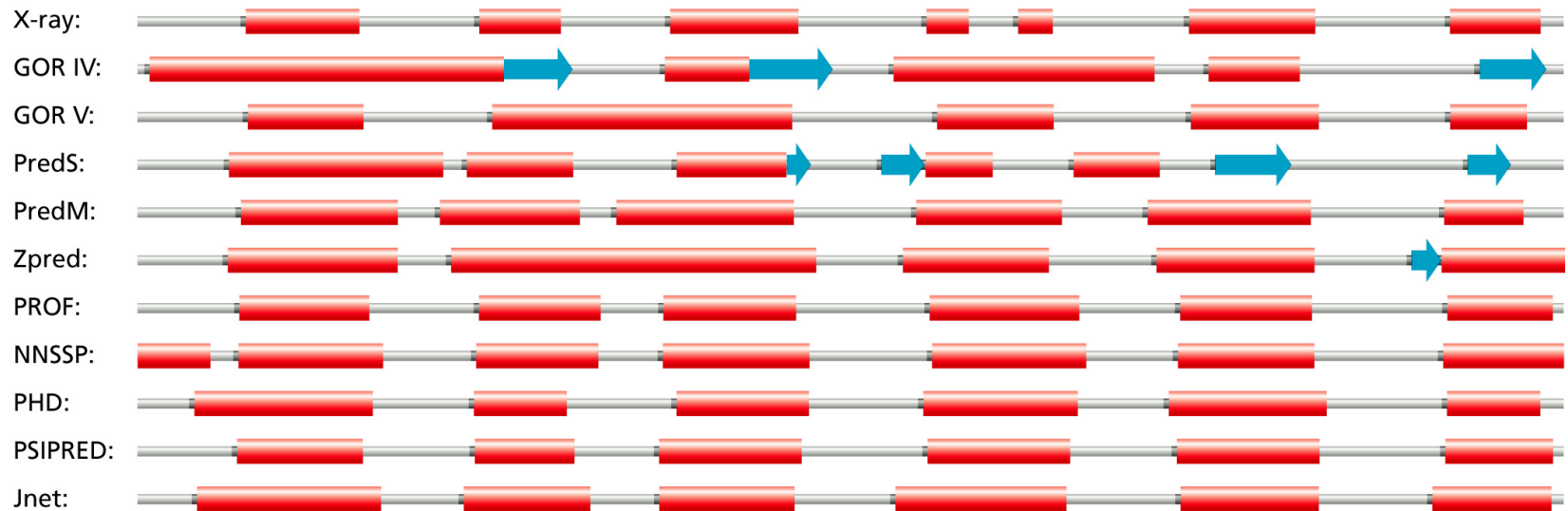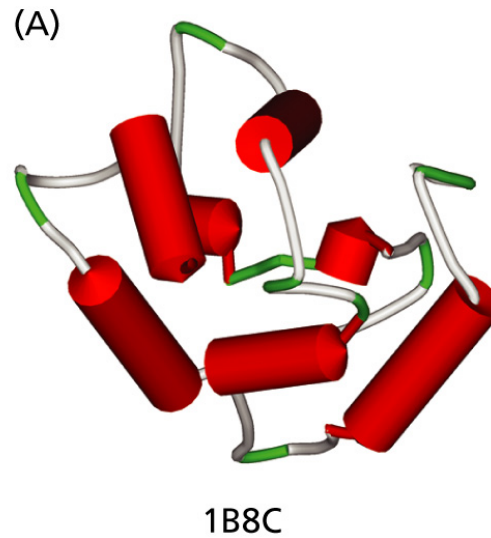
Transmembrane proteins

Membrane proteins are functionally important. For example, the receptors are formed by 1 or more helices spanning the mebrane
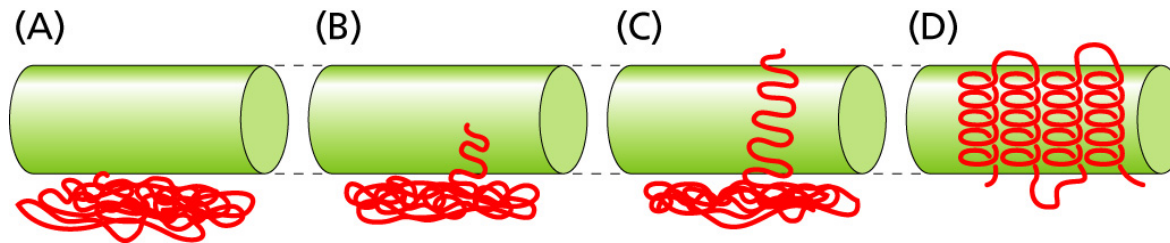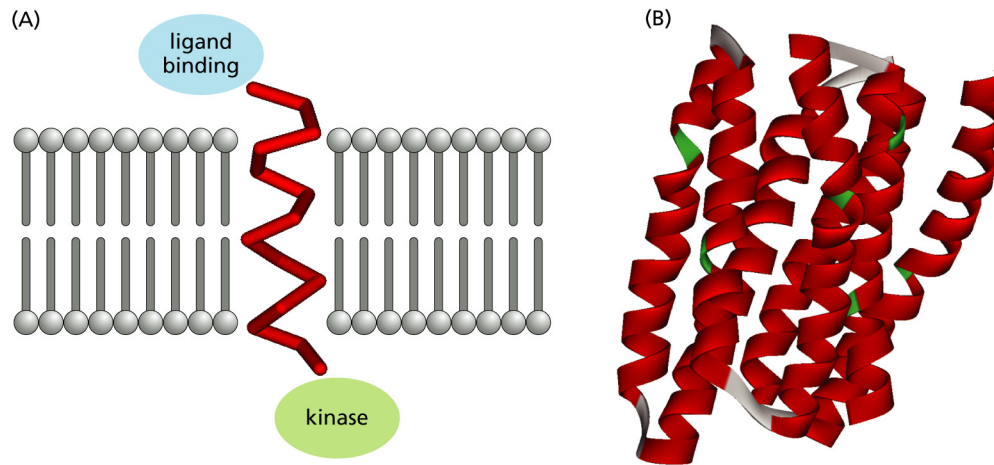
The four main ways in which proteins may be attached to a membrane. A) Attachment by interactions between the protein and the cytosolic face of the lipid bilayer. B) Attachment via an anchor (lipidic or terminals of the protein) that are added post-translationally. C) Transmembrane proteins have part of the protein chain embadded in the lipid bilayer. D) Transmembrane proteins where the protein chain threads back and forth across the mebrane multiple times.

Transmembrane proteins

Helix wheel

(A)

(B)

36 – 63

(C)

208 – 225

● hydrophilic   ● hydrophobic   ● others
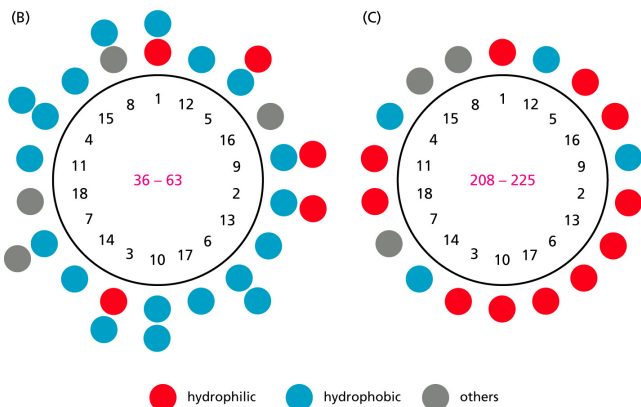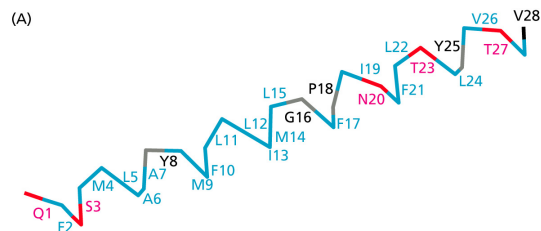
Hydrophobicity diagram

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 p100α 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.

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

Copyright © 2002, W. H. Freeman and Company
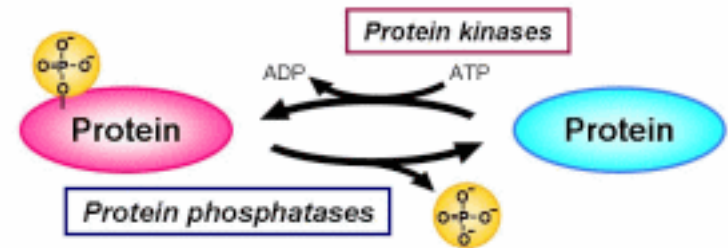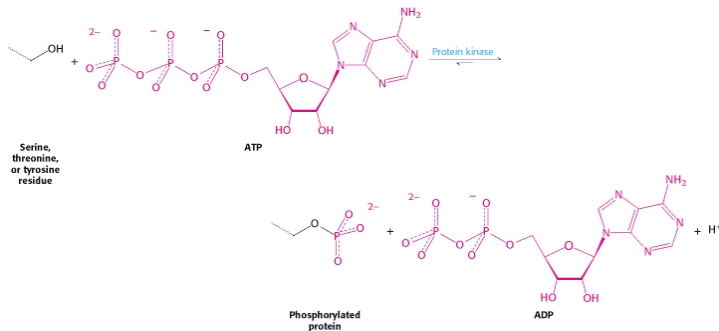
The consensus sequence recognized by protein kinase A is Arg-Arg-X-Ser-Z or Arg-Arg-X-Thr-Z, in which X is a small residue, Z is a large hydrophobic one, and Ser or Thr is the site of phosphorylation. It should be noted that this sequence is not absolutely required.

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
QWERRRTYELVISLIVESYEAHYEAH           <-- 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
_____v_____

test1          13    ELVISLIVE  0.017    .
test1          18    LIVESYEAH  0.942  *S*
_____^_____


            Threonine predictions

Name          Pos    Context    Score  Pred
_____v_____

test1           7    ERRRTYELV  0.921  *T*
_____^_____


            Tyrosine predictions

Name          Pos    Context    Score  Pred
_____v_____

test1           8    RRRTYELVI  0.056    .
test1          19    IVESYEAHY  0.502  *Y*
test1          23    YEAHYEAH-  0.885  *Y*
_____^_____
```


NetPhos 2.0: predicted phosphorylation sites in Sequence