

Dr G.P.S. Raghava

Annotation Methods

- Annotation by homology (BLAST)
 - requires a large, well annotated database of protein sequences

- Annotation by sequence composition
 - simple statistical/mathematical methods
- Annotation by sequence features, profiles or motifs
 - requires sophisticated sequence analysis tools
 - Annotation by Subcellular localization
 - requires computational tools for better subcellular localization prediction.

Annotation by Homology

- Statistically significant sequence matches identified by BLAST searches against GenBank (nr), SWISS-PROT, PIR, ProDom, BLOCKS, KEGG, WIT, Brenda, BIND
- Properties or annotation inferred by name, keywords, features, comments

```
DBSOURCE swissprot: locus MPPB_NEUCR, ...
xrefs (non-sequence databases): ...
InterProIPR001431,...
KEYWORDS Hydrolase; Metalloprotease; Zinc;
Mitochondrion; Transit peptide;
Oxidoreductase; Electron transport;
Respiratory chain.
```

Databases Are Key

Different Levels of Database Annotation

- GenBank (minimal annotation)
- PIR (slightly better annotation)
- SwissProt (even better annotation)
- Organsim-specific DB (best annotation)

Structure Databases

- RCSB-PDB
 - http://www.rcsb.org/pdb/
- MSD
 - http://www.ebi.ac.uk/msd/index.html
- CATH
 - http://www.biochem.ucl.ac.uk/bsm/cath/
- SCOP
 - http://scop.mrc-lmb.cam.ac.uk/scop/

Expression Databases

Metabolism Databases

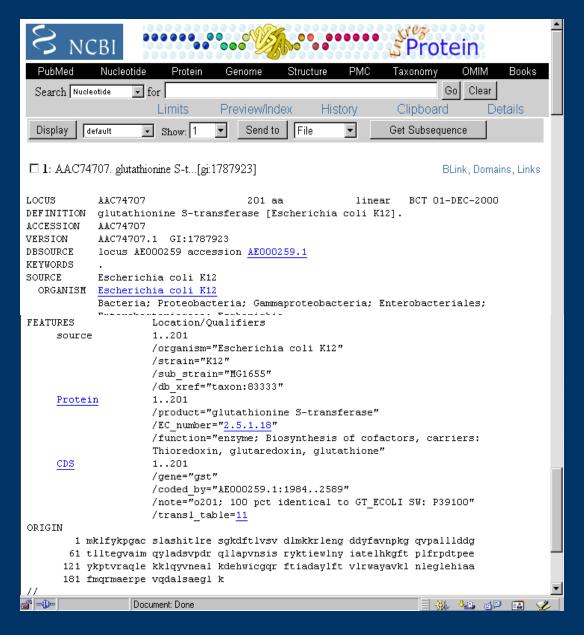
- Swiss 2D Page
 - http://ca.expasy.org/ch2d/
- SMD
 - http://genomewww5.stanford.edu/MicroArr ay/SMD/

- KEGG
 - http://www.genome.ad.jp/ kegg/metabolism.html
- EcoCyc
 - www.ecocyc.org/

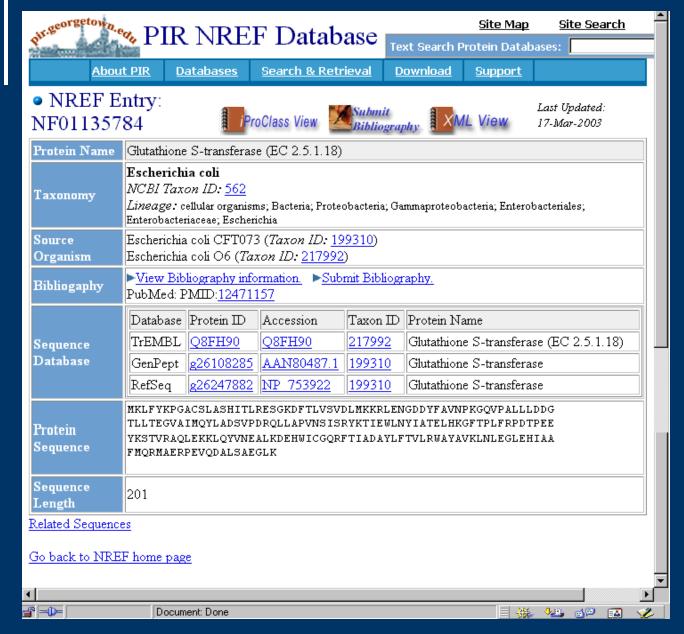
Interaction Databases

- BIND
 - http://www.blueprint.org/bin d/bind.php

GenBank Annotation

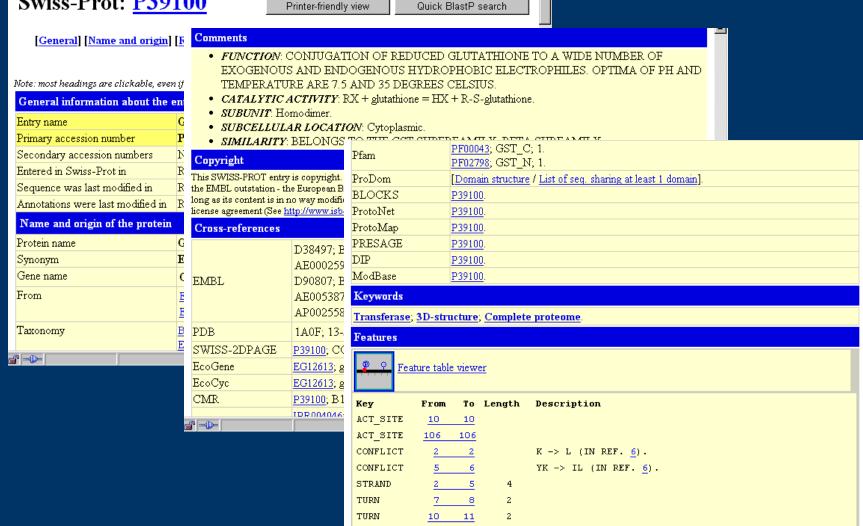


PIR Annotation



Swiss-Prot Annotation



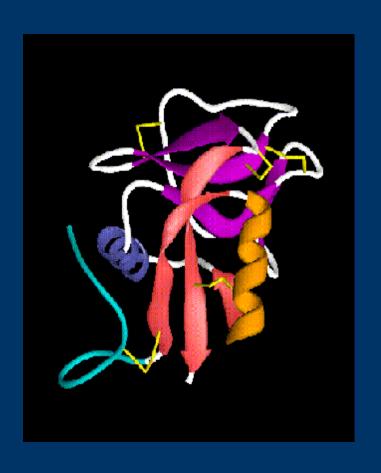


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• • Annotation by Composition

- Molecular Weight
- Isoelectric Point
- UV Absorptivity
- Hydrophobicity



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Feature based annotation

```
Pfam; PF00234; tryp_alpha_amyl; 1.
PROSITE; PS00940; GAMMA_THIONIN; 1.
PROSITE; PS00305; 11S_SEED_STORAGE; 1.

parse

features
```

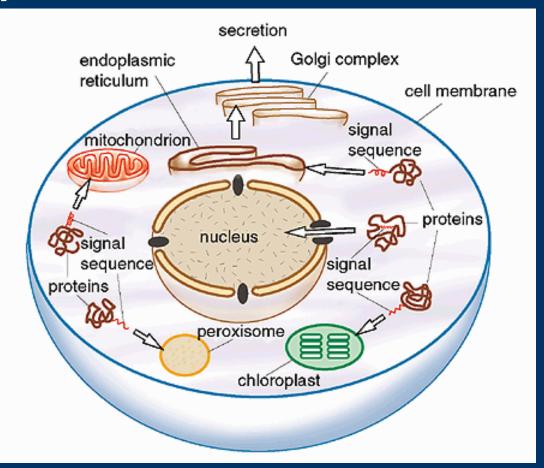
- PROSITE http://www.expasy.ch/
- BLOCKS http://blocks.fhcrc.org/
- DOMO http://www.infobiogen.fr/services/domo/
- PFAM http://pfam.wustl.edu
- PRINTS http://www.biochem.ucl.ac.uk/bsm/dbrowser/PRINTS
- SEQSITE PepTool

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• What is Subcellular Localization?

- Organelles
- Membranes
- Compartments
- Microenvironments



Gene Ontology

Cellular component contains organelles, membranes, cell regions, localized and unlocalized protein complexes

```
□GO:0003673: Gene Ontology (33650)
   ⊕ GO:0008150 : biological process (24768)
   □ ® GO:0005575 : cellular component (17255)
      □ @ GO:0005623 : cell (14268)
         ⊕ o GO:0005627 : ascus (5)
          . 

GO:0030424: axon (2)

⊕ GO:0005933 : bud (58)

         ⊕ @ GO:0000267 : cell fraction (836)
          . @ GO:0030425 : dendrite (3)
         ⊕ GO:0019861 : flagellum (27)

    ⊕ GO:0005622 : intracellular (12771)

         □ ® GO:0016020 : membrane (4511)
             ⊕ © GO:0012505 : endomembrane system (378)
             ⊕ o GO:0009279 : external outer membrane (sensu Gram-negative bacteria) (12)

    ⊕ GO:0019898 : extrinsic membrane protein (58)

             ⊕ @ GO:0019866 : inner membrane (359)

    ⊕ GO:0016021 : integral membrane protein (2268)

             ⊕ o GO:0005765 : lysosomal membrane (62)
             ⊕ o GO:0005740 : mitochondrial membrane (499)
             ⊕ © GO:0019867 : outer membrane (55)
             ⊕ © GO:0005778 : peroxisomal membrane (48)
             ⊕ o GO:0005886 : plasma membrane (2273)
             • 0 GO:0005628 : prospore membrane (4)
             ⊞ o GO:0009579 : thylakoid (101)
             ⊕ o GO:0005774 : vacuolar membrane (81)
```

Subcellular Localization Ontology

- Cellular Components can be instantiated
- Captures spatial relationships
- Maps to GO concepts
- Uses EcoCyc concepts:
 Macromolecule, Reaction, Pathway



Why is Subcellular Localization Important?

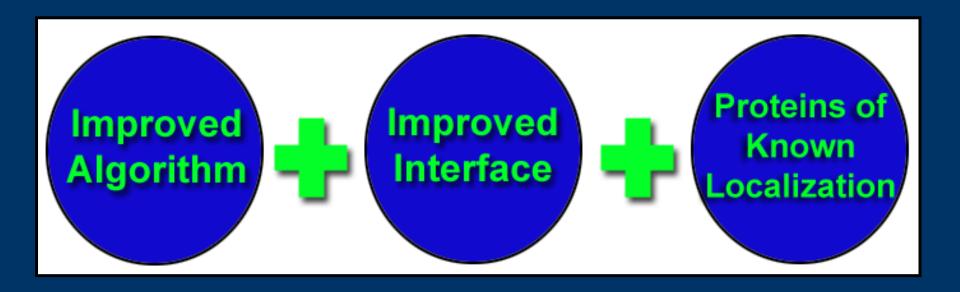
- Function is dependent on context
- Localization is dynamic and changing
- Compartmentalization forms groups which allows for abstraction of concepts (i.e. mitochondria)

Specifying Subcellular Localization: Why is it difficult?

- Biological Context
- Hard to define boundaries
- Dynamic Systems
- Distributions of proteins

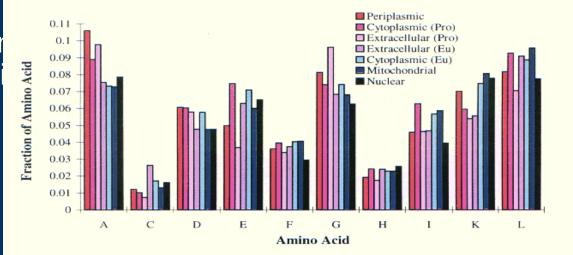
Our solution: Code Name bPSORT

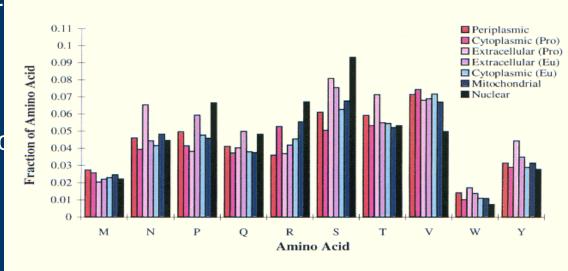




Method 1 - Amino acid composition

- Correlate amir acid compositi to subcellular location
 - Alanine periplasm
 - Glycine extracellular
 - Serine nucleus
 - Leucine mitochondrio







Method 2 - Find signal sequences

- Short stretches of amino acids
- Located at either end of the protein
- Sometime in the middle of the sequence

QuickTime™ and a TIFF (Uncompressed) decompressor are needed to see this picture.



Protein translocation across the ER membrane

- The signal peptide binds to the SRP
- The SRP complex docks on the channel
- The signal peptides is cleaved and the protein is secrete out of the cell

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Method 3 - Combine homologs and NLP

- A lot of well annotated databases
- Sequence alignment to find homologs
- Extract important texts (features) from homologs
- Analyze texts (features) using NLP techniques
- Training predictors based on those features
- Make predictions to new sequences using pre-built predictors



- Composed of separate modules
 - Motif analysis
 - Signal peptide detection
 - Transmembrane domains
- Each predicts one particular location
- Integrate modules by either rulebased system or probabilistic model

Future directions

- Is cross validation convincing?
- Golden datasets for fair evaluation
- Is the prediction result obvious to the users?
 - Transparency
- Is one location per protein enough?
 - Protein transport

Current prediction methods

Eukaryotic localization predictors:

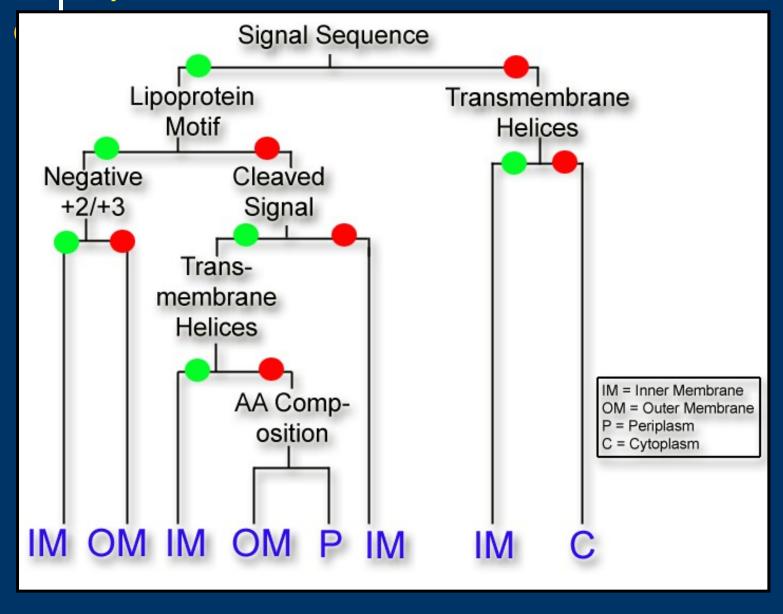
TargetP (Emanuelsson et al, 2001) iPSORT (Bannai et al, 2001) PSORT II (Horton, P. and Nakai, 1997) ESLPred (Bhasin and Raghava, 2004)

- ⇒ Prokaryotic localization predictors:

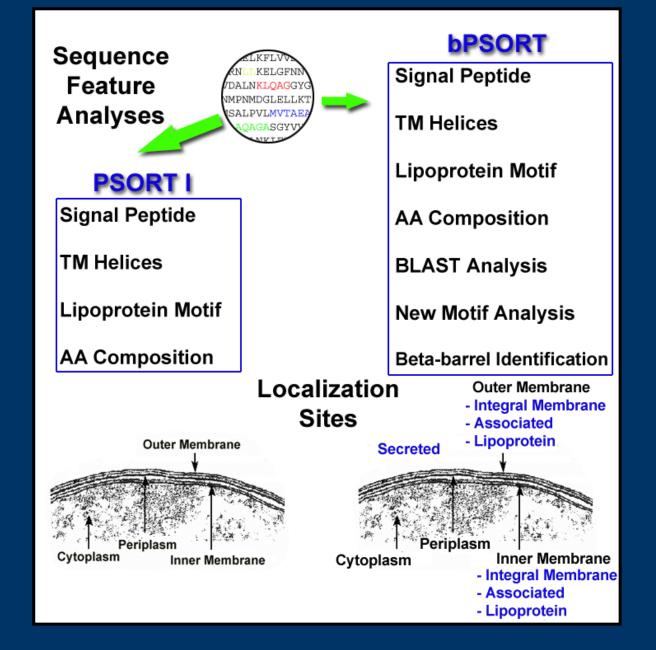
 PSORT I (Nakai, K. and Kanehisa, 1991)

 PSORT-B (Nakai, K. and Kanehisa, 2001)
- ⇒ Eukaryotic and Prokaryotic localization predictors: NNPSL (Reinhardt and Hubbard, 1998) SubLoc (Hau and Sun, 2001)

Current prediction methods – PSORT I



Improved Prediction Algorithm



• • NNPSL

- Use AA composition
- Use neural networks
- Prokaryotic
 - periplasm, cytoplasm and extracellular
 - 81%
- Eukaryotic
 - Extracellular, cytoplasm, mitochondrion and nuclear
 - **66%**

• • • SubLoc

- Use AA composition
- Use SVMs instead of NN in NNPSL
- Same datasets
- Different results
 - 91.4% prokaryotic
 - 79.4% eukaryotic

• • SignalP

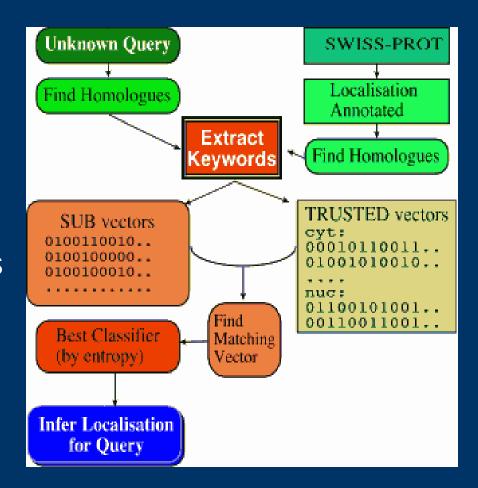
- Current version 3.0 predicts the presence and location of signal peptide cleavage sites
- Based on neural networks in its first version v1.0
- Developed SignalP-HMM in Version 2.0
- Eukaryotic and Gram positive and negative bacteria

TargetP

- Predicts the subcellular location of eukaryotic protein sequences
- Based on the predicted presence of any of the N-terminal short sequences
 - Chloroplast transit peptide (cTP)
 - Mitochondrial targeting peptide (mTP)
 - Secretory pathway signal peptide (SP)

LOCKey

- Look for proteins with known localization in Swiss-Prot
- Construct trusted vectors using reliable homologs
- Match SUB vectors to trusted vectors to make new predictions



Improved Method for Subcellular localization of Eukaryotic protein is Required The PSORT-B is highly accurate method for prediction of subcellular localization of prokaryotic protein. The subcellular localization of eukaryotic proteins is not so accurate due to complexity of proteins.

A highly accurate method for subcellular localization of eukaryotic proteins is of immense importance for better functional annotation of protein.

Attempt for better prediction of subcellular localization of eukaryotic protein

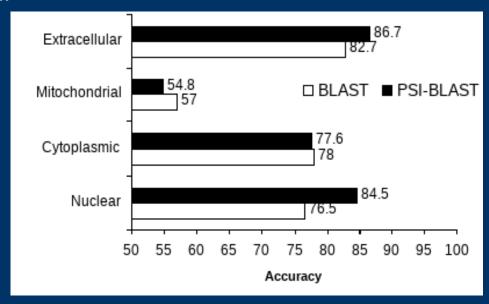
Dataset for classification

- >Experimentally proven proteins
- Complete protein
- Non-redundant proteins(90)

2427

Similarity based prediction of subcellular localization

We have generated a BLAST as well as with PSIBLAST based module for subcellular localization. It performance was evaluated by 5-fold cross-validation.



It proves that PSIBLAST is better as compared to simple BLAST in subcellular localization prediction. Out of 2427 proteins whereas it was only 362 proteins for which no significant hit was found in case of PSI-BLAST.

AS it is proved in past that machine learning techniques are elegant in classifying the biological data. Hau and sun applied the SVM for classification of prokaryotic and eukaryotic proteins and shown that it is better than statistical methods as well as other machine learning techniques such as ANN. For class classification four 1-v-r SVMs were used.

REq: Fixed Pattern length

Q: How to convert the variable length of proteins to fixed length?

Ans: Amino acid composition is most widely used for this purpose.

 $fraction \ of \ aa_{_{i}} = \frac{total \ number \ of \ amino \ acid \ i}{total \ number \ of \ amino \ acids \ of \ protein}$

RESULTS

where i can be any amino acid out of 20 natural amino acids.

Approach	Nuclear		Cytoplasmic		Mitochondrial		Extracellular	
	ACC	MCC	ACC	MCC	ACC	MCC	ACC	MCC
Composition based	86.1	0.73	76.9	0.64	55.5	0.54	76.0	0.76



We have taken in consideration 33 physic-chemical properties for classification like hydrophobocity, hydrophilicity.

Approach	Nuclear		Cytoplasmic		Mitochondrial		Extracellular	
	ACC	МСС	ACC	MCC	ACC	МСС	ACC	МСС
Properties Based	85.6	0.73	74.6	0.64	59.2	0.55	76.6	0.74

The physico-chemical properties-based SVM module predicted subcellular localization of protein with slightly lower accuracy (77.8%) than the amino acid composition based module.

What is lacking in amino acid composition and properties based classification?

Both of the properties provide information about the fraction of residues of particular type and lack information about residue order.

So property that provide information

amino acid composition + Order = More accurate prediction ?

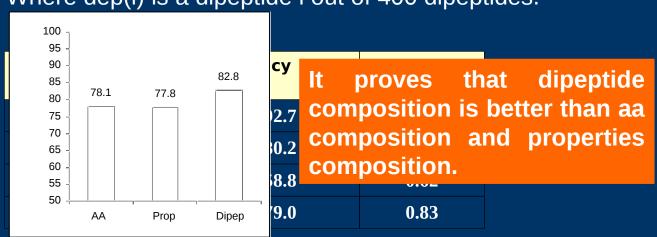
Dipeptide
Tripeptide
Tetrapeptide......

So we have used dipeptide composition for subcellular localization prediction.

fraction of dep (i) =
$$\frac{\text{total number of dep(i)}}{\text{total number of all possible dipeptides}}$$

Where dep(i) is a dipeptide i out of 400 dipeptides.

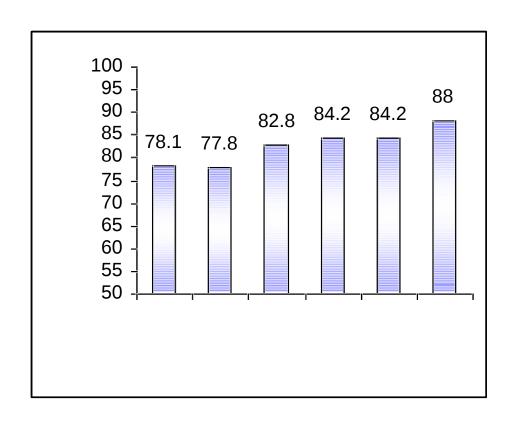
Result



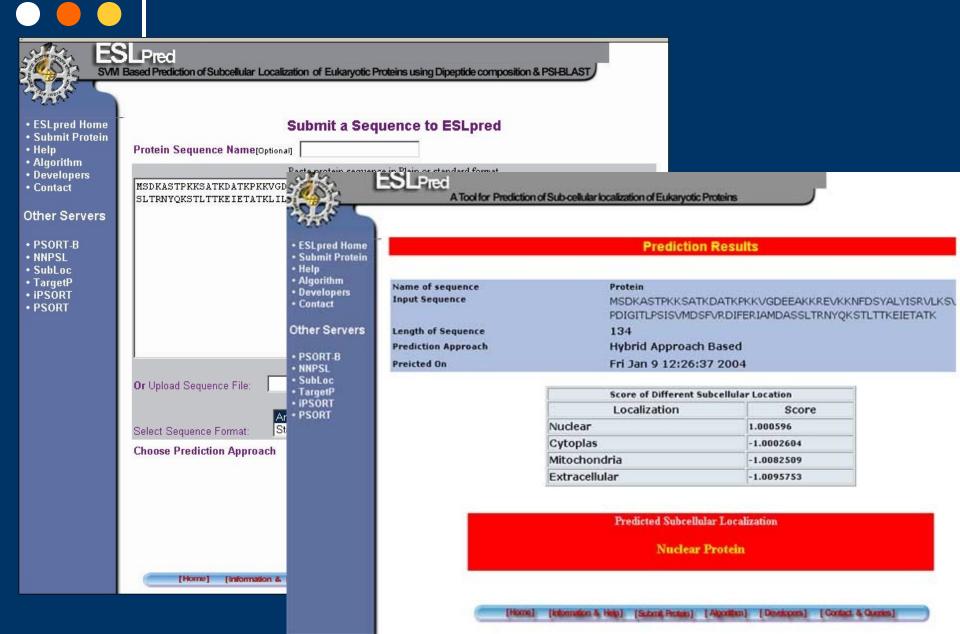
To further improve the accuracy:

Tripeptide composition: SVM fails to train due to complexity of patterns. The pattern of each protein is of 8,000 vectors with lot of noise.

Hybrid approach: Using more then one feature.



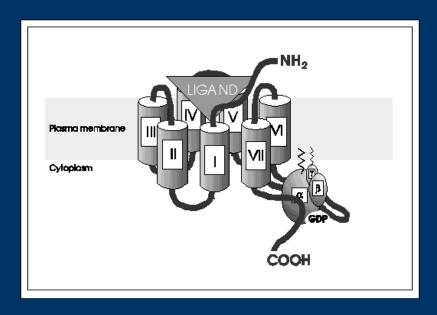
Method based on above results have been implemented online as **ESLPRED**



Functional annotation: Classification

G-Protein Coupled Receptor

- Membrane-bound receptors
- Transducing messages as photons, organic odorants, nucleotides, nucleosides, peptides, lipids and proteins.
- 6 different families
- A very large number of different domains both to bind their ligand and to activate G proteins.



More than 50% of drugs in the market are base on GPCRs due to their major role in signal transduction.

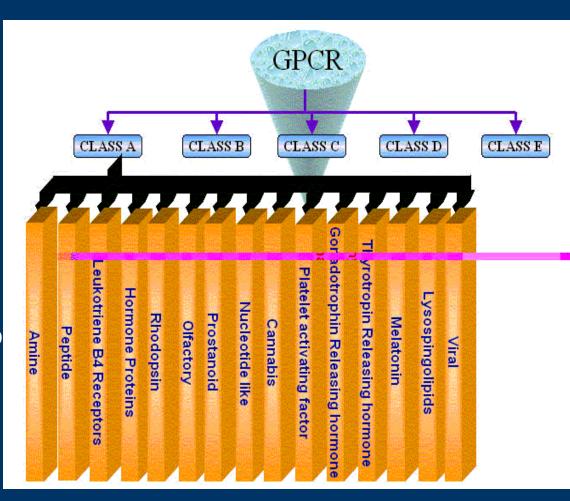


Up to Subfamilies (Part 1)

Types of Receptors of each subfamily (II)

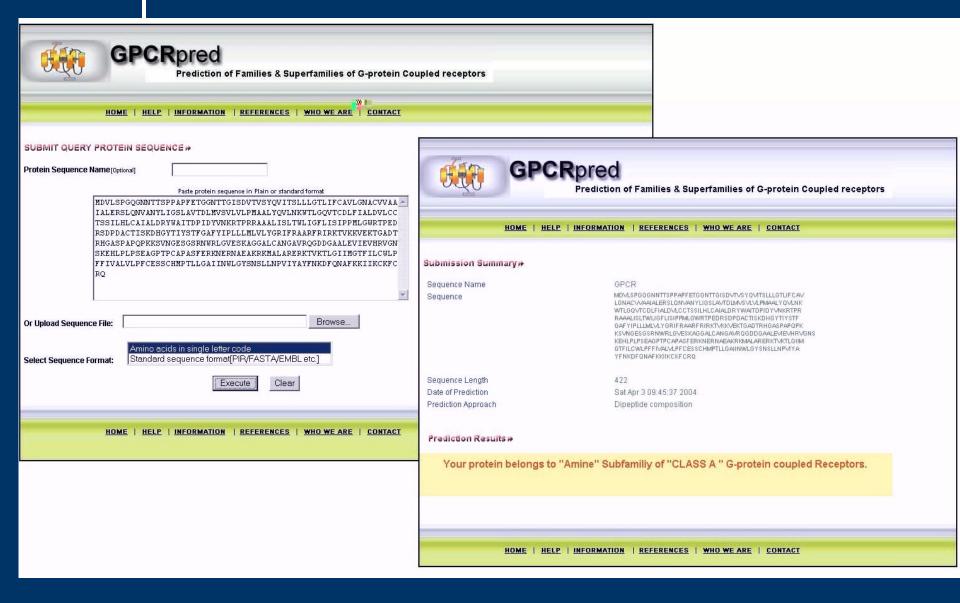
History:

- •Mostly BLAST is used for the classification and recognition of novel GPCRs.
- Motifs search is also used as GPCRs have conserved structure.
- •One SVM based method is also available for classification of GPCRs of Rhodopsin family. (Karchin et al., 2001)









Functional annotation: Classification



- ❖Nuclear receptors are the key transcription factors that regulate crucial gene networks responsible for cell growth, differentiation and homeostasis.
- ❖Potential drug targets for developing therapeutic strategies for diseases like cancer and diabetes.
- classified into seven subfamilies
- **♦** Consist of six distinct regions.

