

For written notes on this lecture, please read chapter 19 of *The Practical Bioinformatician*

CS2220: Introduction to Computational Biology

Lecture 7: Sequence Homology Interpretation

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9 March 2007



Plan

- **Recap of sequence alignment**
- **Guilt by association**
- **Active site/domain discovery**
- **What if no homology of known function is found?**
 - Genome phylogenetic profiling
 - Protfun
 - SVM-Pairwise
 - Protein-protein interactions
- **Key mutation site discovery**

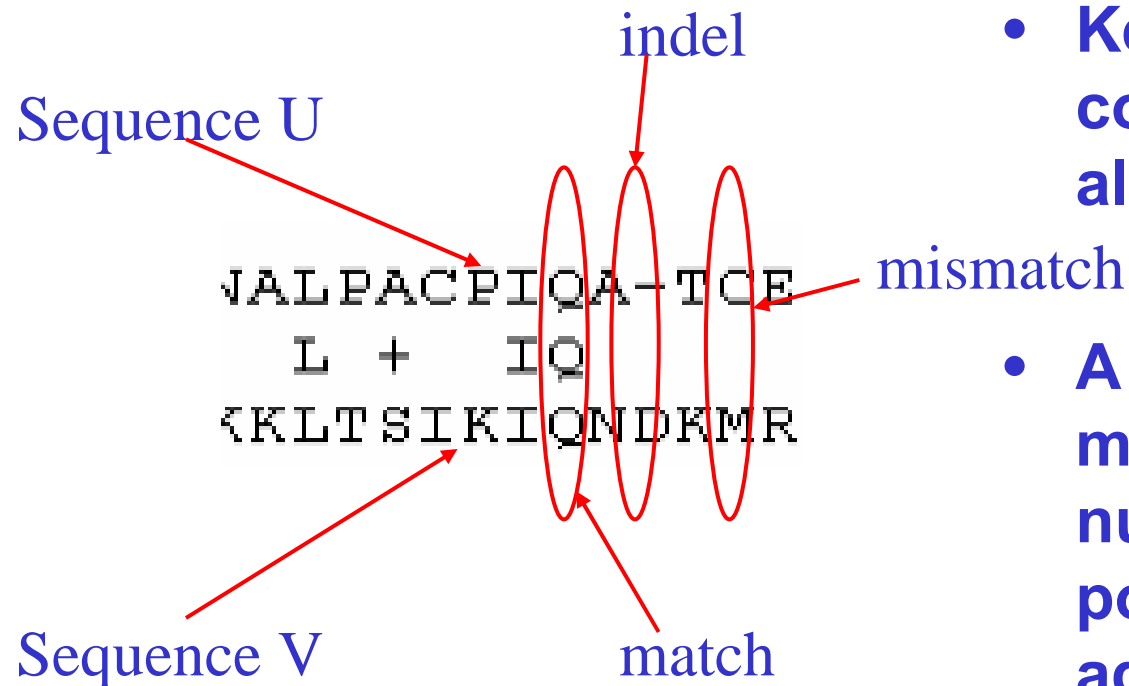
Very Brief Recap of Sequence Comparison/Alignment



Motivations for Sequence Comparison

- **DNA is blue print for living organisms**
 - ⇒ **Evolution is related to changes in DNA**
 - ⇒ **By comparing DNA sequences we can infer evolutionary relationships between the sequences w/o knowledge of the evolutionary events themselves**
- **Foundation for inferring function, active site, and key mutations**

Sequence Alignment



- Key aspect of seq comparison is seq alignment
- A seq alignment maximizes the number of positions that are in agreement in two sequences

Sequence Alignment: Poor Example

- Poor seq alignment shows few matched positions
 ⇒ The two proteins are not likely to be homologous

Alignment by FASTA of the sequences of amicyanin and domain 1 of ascorbate oxidase

	60	70	80	90	100
Amicyanin	MPHNVHFFVAGVLGEAALKGPMKKEQAYSLTFTEAGTYDYHCTPHPFMRGKVVVE				
		
Ascorbate Oxidase	ILQRGTFWADGTASISQCAINPGETFFYNFTVDNPGTFFYHGHLMQRSAGLYGSLI				
	70	80	90	100	110 120

No obvious match between
 Amicyanin and Ascorbate Oxidase

Sequence Alignment: Good Example

- Good alignment usually has clusters of extensive matched positions
- ⇒ The two proteins are likely to be homologous

```
□ >gil13476732|ref|NP\_108301.1| unknown protein [Mesorhizobium loti]  
gil14027493|dbj|BAB53762.1| unknown protein [Mesorhizobium loti]  
Length = 105
```

Score = 105 bits (262), Expect = 1e-22

Identities = 61/106 (57%), Positives = 73/106 (68%), Gaps = 1/106 (0%)

```
Query: 1   MKPGRLASIALAIIFLPMAVPAHAATIEITMENLVISPTEVSAKVGDTIRWVNKDVF AHT 60  
          MK G L  ++      MA PA AATIE+T++ LV SP  V AKVGDTI WVN DV AHT  
Sbjct: 1   MKAGALIRLSWLAALALMAAPAAAATIEVTIDKLVFSPATVEAKVGDTIEWVNNDVVAHT 60
```

good match between
Amicyanin and unknown M. loti protein

Multiple Alignment: An Example

- Multiple seq alignment maximizes number of positions in agreement across several seqs
- seqs belonging to same “family” usually have more conserved positions in a multiple seq alignment

```

gi|126467|      FHFTSWPDFGVPFTPIGMLKFLKKVKACNP--QYAGAIIVHCSAGVGRTGTFVVIDAMLD
gi|2499753      FHFTGWPDHGVVPYHATGLLSFIRRVKLSNP--PSAGPIVHCSAGAGRTGTCYIVIDIMLD
gi|462550|      YHYTQWPDIMGVPEYALPVLTFVRRSSAARM--PETGPVIVHCSAGVGRTGTYYIVIDSMLQ
gi|2499751      FHFTSWPDHGVDPDTTDLLINFRYLVRDYMKQSPPEPILVHCSAGVGRTGTFIAIDRLIY
gi|1709906      FQFTAUPDHGVPEHPTPFLAFLRRVKTCNP--PDAGPMIVHCSAGVGRTGCFIVIDAMLE
gi|126471|      LHFTSWPDFGVPFTPIGMLKFLKKVKTLNP--VHAGPIVHCSAGVGRTGTFIVIDAMMA
gi|548626|      FHFTGWPDHGVVPYHATGLLSFIRRVKLSNP--PSAGPIVHCSAGAGRTGTCYIVIDIMLD
gi|131570|      FHFTGWPDHGVVPYHATGLLGFRVQVKS KSP--PNAGPLVHCSAGAGRTGCFIVIDIMLD
gi|2144715      FHFTSWPDHGVDPDTTDLLINFRYLVRDYMKQSPPEPILVHCSAGVGRTGTFIAIDRLIY
                ..*  ***  ***      .  *      .....*  *****...  **  ..
  
```

Conserved sites

Application of Sequence Comparison: Guilt-by-Association



A protein is a ...

- A protein is a large complex molecule made up of one or more chains of amino acids
- Protein performs a wide variety of activities in the cell



Function Assignment to Protein Sequence

SPSTNRKYPPLPVDKLEEEINRRMADDNKLFREEFNALPACPIQATCEAASKEENKEKNR
YVNILPYDHSRVHLTPVEGVPSDYINASFINGYQEKNKFIAAQGPKEETVNDFWRMIWE
QNTATIVMTNLKERKECKCAQYWPDQGCWTYGNVRVSVEDVTVLVDYTVRKFCIQQVGD
VTNRKPQRLITQFHFTSWPDFGVPTPIGMLKFLKKVKACNPQYAGAIVVHCSAGVGRGTG
TFVVIDAMLDMMSERKVDVYGFVSRIRAQRCQMVQTD MQYVFIYQALLEHYLYGDTELE
VT

- How do we attempt to assign a function to a new protein sequence?

Guilt-by-Association

- Compare the target sequence T with sequences S_1, \dots, S_n of known function in a database
- Determine which ones amongst S_1, \dots, S_n are the mostly likely homologs of T
- Then assign to T the same function as these homologs
- Finally, confirm with suitable wet experiments

Guilt-by-Association

Compare T with seqs of known function in a db

Poor Sequence Alignment

- Poor seq alignment shows few matched positions
 \Rightarrow The two proteins are not likely to be homologous

Alignment by FASTA of the sequences of amicyanin and domain 1 of ascorbate oxidase

	60	70	80	90	100
Amicyanin	MPHNVHFVAGVLGEAALKGPMKKEQAYS	LTFTTEAGTYDYHCTPHFFMRGKVVV			
	
Ascorbate Oxidase	ILQRGTPWADGTASISQCAINPGETFF	YFNFTVDNPGTFFYHGHLMQORSAGLYG			
	70	80	90	100	110

No obvious match between Amicyanin and Ascorbate Oxidase

Discard this function as a candidate

Good Sequence Alignment

- Good alignment usually has clusters of extensive matched positions
 \Rightarrow The two proteins are likely to be homologous

```
>gi113476732|ref|NP_108301.1| unknown protein [Mesorhizobium loti]
gi114027493|db|BAE53762.1| unknown protein [Mesorhizobium loti]
Length = 105

Score = 105 bits (262), Expect = 1e-22
Identities = 61/106 (57%), Positives = 73/106 (68%), Gaps = 1/106 (0%)

Query: 1 MKPORLASIALAIIFLPMVAFARAATIEITMENLVISPTIEVSAKVVDITKPVNKKVFAHT 60
      MK G L ++ MA PA AATIE+T++ LV SP V AKVGDIT WVN DV AHT
Sbjct: 1 MKAGALIELSLAALALMAFAAAATIEVTIDKLVPSPATVEAKVGDITKPVNKKVFAHT 60
```

good match between Amicyanin and unknown M. loti protein

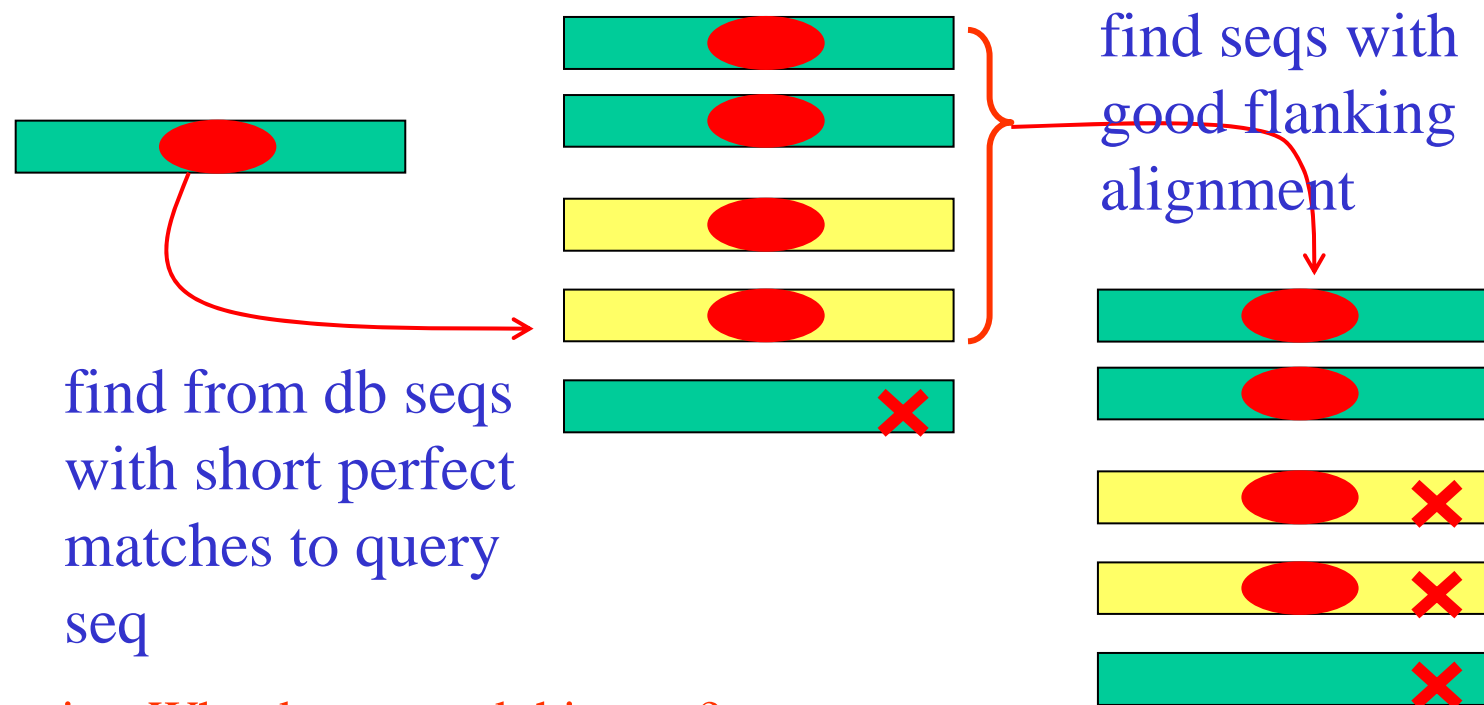
Assign to T same function as homologs

Confirm with suitable wet experiments

BLAST: How It Works

Altschul et al., *JMB*, 215:403--410, 1990

- **BLAST is one of the most popular tool for doing “guilt-by-association” sequence homology search**



Exercise: Why do we need this step?

Homologs obtained by BLAST

Sequences producing significant alignments:	Score (bits)	E Value
gi 14193729 gb AAK56109.1 AF332081.1 protein tyrosin phosph...	621 L	e-177
gi 126467 sp P18433 PTRA_HUMAN Protein-tyrosine phosphatase...	621 L	e-177
gi 4506303 ref NP_002827.1 protein tyrosine phosphatase, r...	621 L	e-176
gi 227294 prf 1701300A protein Tyr phosphatase	620	e-176
gi 18450369 ref NP_543030.1 protein tyrosine phosphatase, ...	621 L	e-176
gi 32067 emb CAA37447.1 tyrosine phosphatase precursor [Ho...	611 L	e-176
gi 285113 pir JC1285 protein-tyrosine-phosphatase (EC 3.1....	619	e-176
gi 6981446 ref NP_036895.1 protein tyrosine phosphatase, r...	611 L	e-176
gi 2098414 pdb 1YFO A Chain A, Receptor Protein Tyrosine Ph...	61 S	e-174
gi 32313 emb CAA38662.1 protein-tyrosine phosphatase [Homo...	61 L	e-174
gi 450583 gb AAB04150.1 protein tyrosine phosphatase >gi 4...	605	e-172
gi 6679557 ref NP_033006.1 protein tyrosine phosphatase, r...	601 L	e-172
gi 483922 gb AAA17990.1 protein tyrosine phosphatase alpha	599	e-170

- Thus our example sequence could be a protein tyrosine phosphatase α (PTP α)

Example Alignment with PTP α

Score = 632 bits (1629), Expect = e-180
Identities = 294/302 (97%), Positives = 294/302 (97%)

```

Query: 1   SPSTNRKYPPLPVDKLEEEINRRMADDNKLFREEFNALPACP IQATCEAASXXXXXXXXXXR 60
          SPSTNRKYPPLPVDKLEEEINRRMADDNKLFREEFNALPACP IQATCEAAS      R
Sbjct: 202 SPSTNRKYPPLPVDKLEEEINRRMADDNKLFREEFNALPACP IQATCEAASKEENKEKNR 261

Query: 61  YVNILPYDHSRVHLTPVEGVFSDYINASF INGYQEKKNF IAAQGPKEETVNDFWRMIWE 120
          YVNILPYDHSRVHLTPVEGVFSDYINASF INGYQEKKNF IAAQGPKEETVNDFWRMIWE
Sbjct: 262 YVNILPYDHSRVHLTPVEGVFSDYINASF INGYQEKKNF IAAQGPKEETVNDFWRMIWE 321

Query: 121 QNTATIVMVTNLKERKECKCAQYWPDQGCW TYGNVRVSVEDVTVLVDYTVRKFC IQQVGD 180
          QNTATIVMVTNLKERKECKCAQYWPDQGCW TYGNVRVSVEDVTVLVDYTVRKFC IQQVGD
Sbjct: 322 QNTATIVMVTNLKERKECKCAQYWPDQGCW TYGNVRVSVEDVTVLVDYTVRKFC IQQVGD 381

Query: 181 VTNRKPQRLITQFHFTSWPDFGV PFTP IGMLKFLKKVKACNPQYAGAI VVHCSAGVGRTG 240
          VTNRKPQRLITQFHFTSWPDFGV PFTP IGMLKFLKKVKACNPQYAGAI VVHCSAGVGRTG
Sbjct: 382 VTNRKPQRLITQFHFTSWPDFGV PFTP IGMLKFLKKVKACNPQYAGAI VVHCSAGVGRTG 441

Query: 241 TFVVIDAMLDMMHSEKVDVYGFVSR IRAQRQCMVQ TDMQYVF IYQALLEHYLYGDTELE 300
          TFVVIDAMLDMMHSEKVDVYGFVSR IRAQRQCMVQ TDMQYVF IYQALLEHYLYGDTELE
Sbjct: 442 TFVVIDAMLDMMHSEKVDVYGFVSR IRAQRQCMVQ TDMQYVF IYQALLEHYLYGDTELE 501

```


Guilt-by-Association: Caveats

- **Ensure that the effect of database size has been accounted for**
- **Ensure that the function of the homology is not derived via invalid “transitive assignment”**
- **Ensure that the target sequence has all the key features associated with the function, e.g., active site and/or domain**

Law of Large Numbers

- Suppose you are in a room with 365 other people
- Q: What is the prob that a specific person in the room has the same birthday as you?
- A: $1/365 = 0.3\%$
- Q: What is the prob that there is a person in the room having the same birthday as you?
- A: $1 - (364/365)^{365} = 63\%$
- Q: What is the prob that there are two persons in the room having the same birthday?
- A: 100%

Interpretation of P-value

- Seq. comparison progs, e.g. BLAST, often associate a P-value to each hit
 - P-value is interpreted as prob that a random seq has an equally good alignment
 - Suppose the P-value of an alignment is 10^{-6}
 - If database has 10^7 seqs, then you expect $10^7 * 10^{-6} = 10$ seqs in it that give an equally good alignment
- ⇒ Need to correct for database size if your seq comparison prog does not do that!

Exercise: Name a commonly used method for correcting p-value for a situation like this

Lightning Does Strike Twice!

- **Roy Sullivan, a former park ranger from Virginia, was struck by lightning 7 times**
 - 1942 (lost big-toe nail)
 - 1969 (lost eyebrows)
 - 1970 (left shoulder seared)
 - 1972 (hair set on fire)
 - 1973 (hair set on fire & legs seared)
 - 1976 (ankle injured)
 - 1977 (chest & stomach burned)
- **September 1983, he committed suicide**



Cartoon: Ron Hipschman
Data: David Hand

Effect of Seq Compositional Bias

- One fourth of all residues in protein seqs occur in regions with biased amino acid composition
 - Alignments of two such regions achieves high score purely due to segment composition
- ⇒ While it is worth noting that two proteins contain similar low complexity regions, they are best excluded when constructing alignments
- E.g., by default, BLAST employs the SEG algo to filter low complexity regions from proteins before executing a search

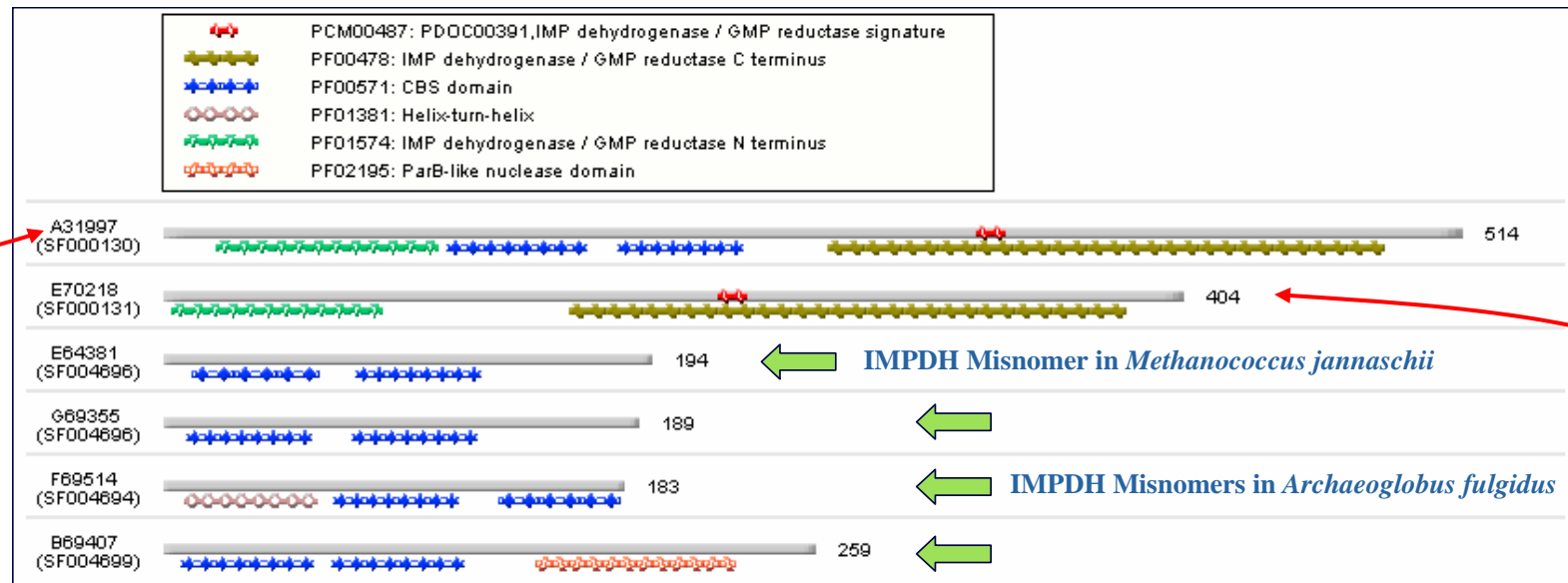
Source: NCBI

Examples of Invalid Function Assignment: The IMP Dehydrogenases (IMPDH)

18 entries were found

ID	Organism	PIR	Swiss-Prot/TrEMBL	RefSeq/GenPept
NF00181857	Methanococcus jannaschii	E64381 conserved hypothetical protein MJ0653	Y653_METJ1A Hypothetical protein MJ0653	g1592300 inosine-5'-monophosphate dehydrogenase (guaB) NP_247637 inosine-5'-monophosphate dehydrogenase (guaB)
NF00187788	Archaeoglobus fulgidus	G69355 MJ0653 homolog AF0847 <i>ALT_NAMES</i> : inosine-monophosphate dehydrogenase (guaB-1) homolog [misnomer]	O29411 INOSINE MONOPHOSPHATE DEHYDROGENASE (GUAB-1)	g2649754 inosine monophosphate dehydrogenase (guaB-1) NP_069681 inosine monophosphate dehydrogenase (guaB-1)
NF00188267	Archaeoglobus fulgidus	F69514 yhcV homolog 2 <i>ALT_NAMES</i> : inosine-monophosphate dehydrogenase (guaB-2) homolog [misnomer]	O28162 INOSINE MONOPHOSPHATE DEHYDROGENASE (GUAB-2)	g2648410 inosine monophosphate dehydrogenase (guaB-2) NP_070943 inosine monophosphate dehydrogenase (guaB-2)
NF00188697	Archaeoglobus fulgidus	A partial list of IMP dehydrogenase misnomers in complete genomes remaining in some public databases		inosine-5'-monophosphate dehydrogenase related protein V
NF00197776	Thermoplasma volcanum			inosine-5'-monophosphate dehydrogenase related protein VII
NF00414709	Methanothermobacter thermautotrophicus			inosine-5'-monophosphate dehydrogenase related protein IX
NF00414811	Methanothermobacter thermautotrophicus	D69035 MJ1232 protein homolog MTH126 <i>ALT_NAMES</i> : inosine-5'-monophosphate dehydrogenase related protein VII [misnomer]	O27294 INOSINE-5'-MONOPHOSPHATE DEHYDROGENASE RELATED PROTEIN V	g2621166 inosine-5'-monophosphate dehydrogenase related protein VII NP_275269 inosine-5'-monophosphate dehydrogenase related protein VII
NF00414837	Methanothermobacter thermautotrophicus	H69232 MJ1225-related protein MTH992 <i>ALT_NAMES</i> : inosine-5'-monophosphate dehydrogenase related protein IX [misnomer]	O27073 INOSINE-5'-MONOPHOSPHATE DEHYDROGENASE RELATED PROTEIN IX	g2622093 inosine-5'-monophosphate dehydrogenase related protein IX NP_276127 inosine-5'-monophosphate dehydrogenase related protein IX
NF00414969	Methanothermobacter thermautotrophicus	B69077 yhcV homolog 2 <i>ALT_NAMES</i> : inosine-monophosphate dehydrogenase related protein X [misnomer]	O27616 INOSINE-5'-MONOPHOSPHATE DEHYDROGENASE RELATED PROTEIN X	g2622697 inosine-5'-monophosphate dehydrogenase related protein X NP_276687 inosine-5'-monophosphate dehydrogenase related protein X










IMPDH Domain Structure



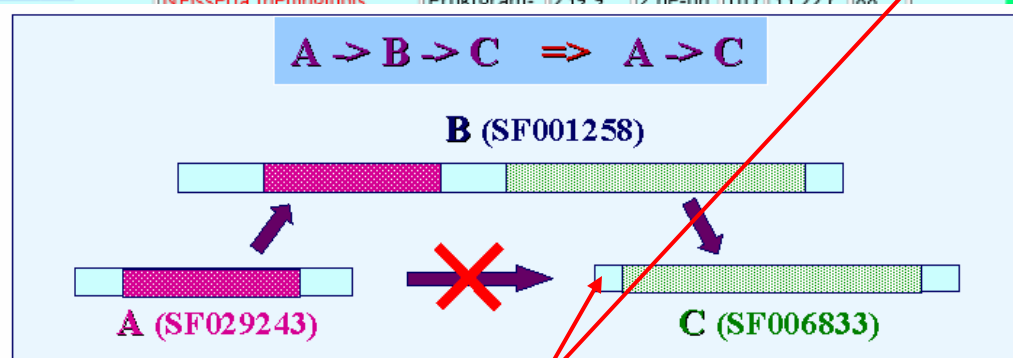
- Typical IMPDHs have 2 IMPDH domains that form the catalytic core and 2 CBS domains.
- A less common but functional IMPDH (E70218) lacks the CBS domains.
- Misnomers show similarity to the CBS domains

Invalid Transitive Assignment

Root of invalid transitive assignment

B →	<input type="checkbox"/> H70468	SF001258	051440	phosphoribosyl-AMP cyclohydrolase (EC 3.5.4.19) / phosphoribosyl-ATP pyrophosphatase (EC 3.6.1.31) [similarity]	Aquifex aeolicus	Prok/other	594.3	4.8e-26	205	39.086	197	
	<input type="checkbox"/> S76963	SF001258	039935	phosphoribosyl-AMP cyclohydrolase (EC 3.5.4.19) / phosphoribosyl-ATP pyrophosphatase (EC 3.6.1.31) [similarity]	Synechocystis sp.	Prok/gram-	557.0	5.7e-24	230	39.175	194	
	<input type="checkbox"/> T35073	SF029243	005738	probable phosphoribosyl-AMP cyclohydrolase	Streptomyces coelicolor	Prok/gram+	399.3	3.5e-15	128	42.157	102	
	<input type="checkbox"/> S53349	SF001257	001188	phosphoribosyl-AMP cyclohydrolase (EC 3.5.4.19) / phosphoribosyl-ATP pyrophosphatase (EC 3.6.1.31) / histidinol dehydrogenase (EC 1.1.1.23)	Saccharomyces cerevisiae	Euk/fungi	384.1	2.5e-14	799	31.863	204	
A →	<input type="checkbox"/> E69493	SF029243	005738	phosphoribosyl-AMP cyclohydrolase (EC 3.5.4.19) [similarity]	Archaeoglobus fulgidus	Archae	396.8	4.8e-15	108	47.778	90	
C →	<input type="checkbox"/> G64337	SF006833	030827	phosphoribosyl-ATP pyrophosphatase (EC 3.6.1.31) [similarity]	Methanococcus jannaschii	Archae	246.9	1.1e-06	95	36.842	95	
	<input type="checkbox"/> D81178	SF006833	101491	phosphoribosyl-ATP pyrophosphatase (EC 3.6.1.31) NMB0603 [similarity]	Neisseria meningitidis	Prok/gram-	239.9	2.6e-06	107	35.227	88	
	<input type="checkbox"/> G81925	SF006833	101491	phosphoribosyl-ATP pyrophosphatase (EC 3.6.1.31) NMA0807 [similarity]								
	<input type="checkbox"/> S51513	SF001257	001188	phosphoribosyl-AMP cyclohydrolase (EC 3.5.4.19) / phosphoribosyl-ATP pyrophosphatase (EC 3.6.1.31) / histidinol dehydrogenase (EC 1.1.1.23)								

Mis-assignment of function

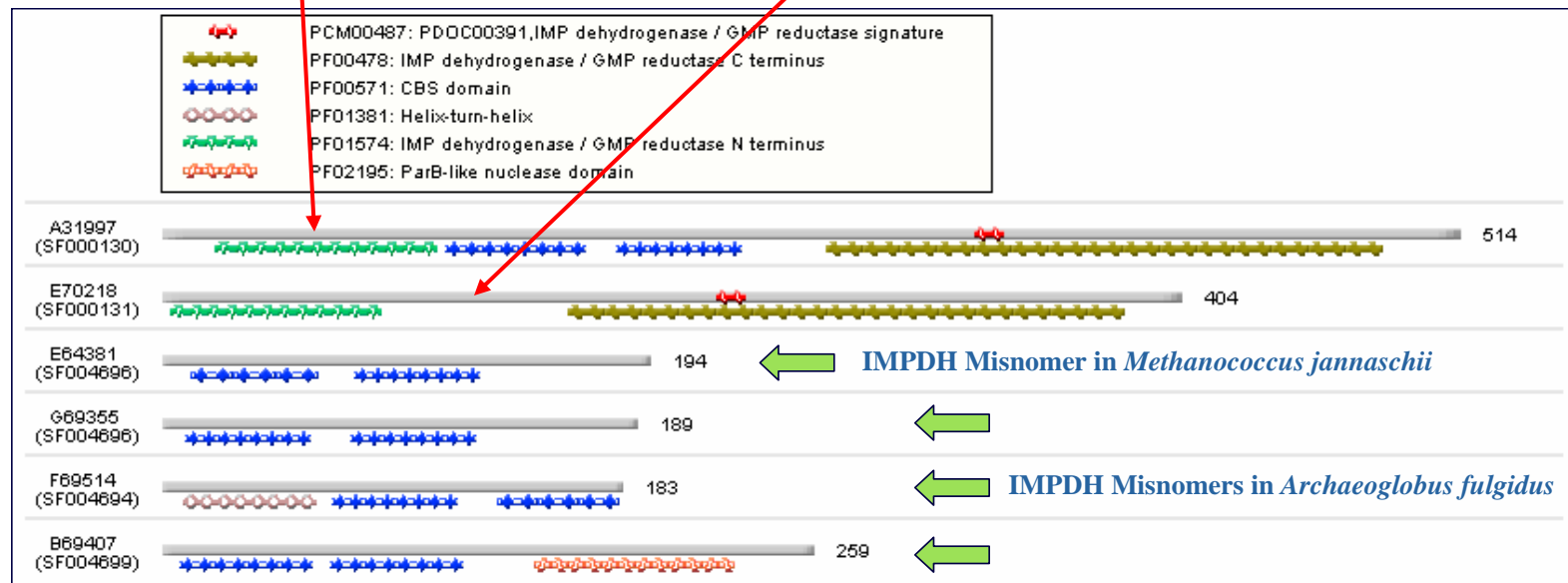


No IMPDH domain

Emerging Pattern

Typical IMPDH

Functional IMPDH w/o CBS



- Most IMPDHs have 2 IMPDH and 2 CBS domains
 - Some IMPDH (E70218) lacks CBS domains
- ⇒ IMPDH domain is the emerging pattern

Application of Sequence Comparison: Active Site/Domain Discovery



Discover Active Site and/or Domain

- **How to discover the active site and/or domain of a function in the first place?**
 - Multiple alignment of homologous seqs
 - Determine conserved positions
 - ⇒ Emerging patterns relative to background
 - ⇒ Candidate active sites and/or domains
- **Easier if sequences of distance homologs are used**

Exercise: Why?

Multiple Alignment of PTPs

```
gi|126467|      FHFTSWPDFGVPFTPIGMLKFLKKVKACNP--QYAGAIVVHCSAGVGRTGTFVVIDAMLD
gi|2499753      FHFTGWPDHGVPPYHATGLLSFIRRVKLSNP--PSAGPIVVHCSAGAGRTGCIYIVIDIMLD
gi|462550|      YHYTQWPDMGVPEYALPVLTFVRRSSAARM--PETGPVLVHCSAGVGRTGTIYIVIDSMLQ
gi|2499751      FHFTSWPDHGVPDTTDLLINFRYLVRDYMKQSPPEPILVHCSAGVGRTGTFIAIDRLIY
gi|1709906      FQFTA WPDHGVPEHPTPFLAFLRRVKTCNP--PDAGPMVVHCSAGVGRTGCFIVIDAMLE
gi|126471|      LHFTSWPDFGVPFTPIGMLKFLKKVKTLNP--VHAGPIVVHCSAGVGRTGTFIVIDAMMA
gi|548626|      FHFTGWPDHGVPPYHATGLLSFIRRVKLSNP--PSAGPIVVHCSAGAGRTGCIYIVIDIMLD
gi|131570|      FHFTGWPDHGVPPYHATGLLG FVRQVKS KSP--PNAGPLVVHCSAGAGRTGCFIVIDIMLD
gi|2144715      FHFTSWPDHGVPDTTDLLINFRYLVRDYMKQSPPEPILVHCSAGVGRTGTFIAIDRLIY
                ..*  ***  ***          .  *                      ..*****  ****...  **  ..
```

- Notice the PTPs agree with each other on some positions more than other positions
 - These positions are more imp't wrt PTPs
 - Else they wouldn't be conserved by evolution
- ⇒ They are candidate active sites

Guilt-by-Association: What if no homolog of known function is found?

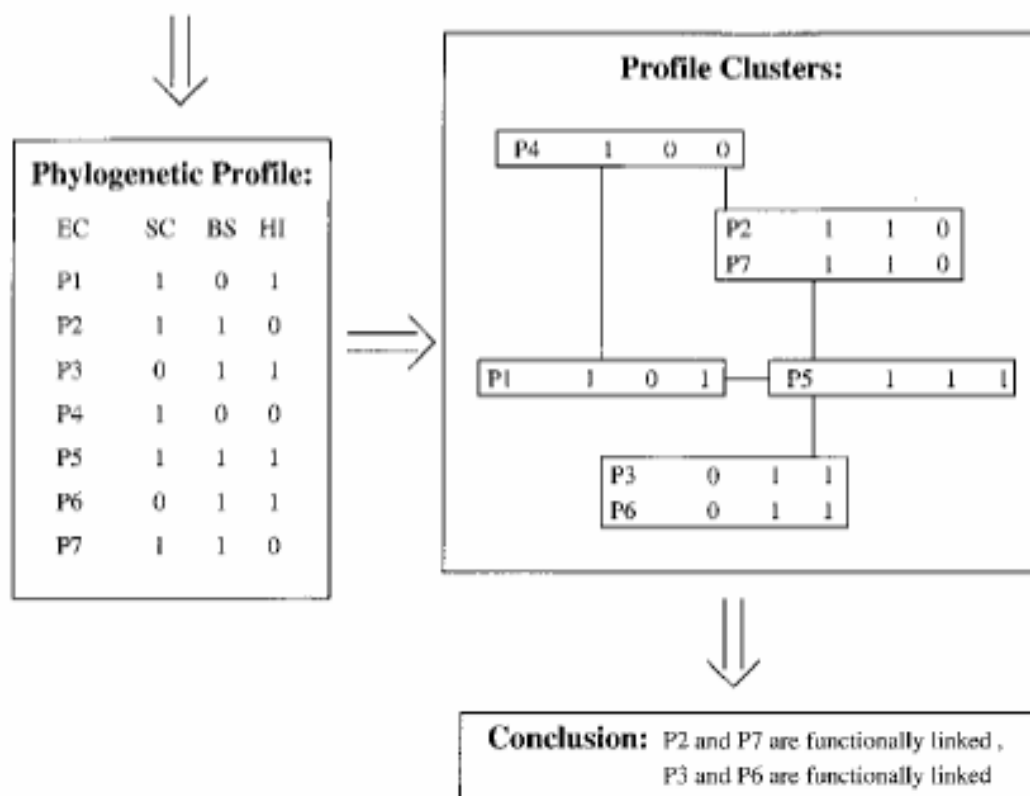
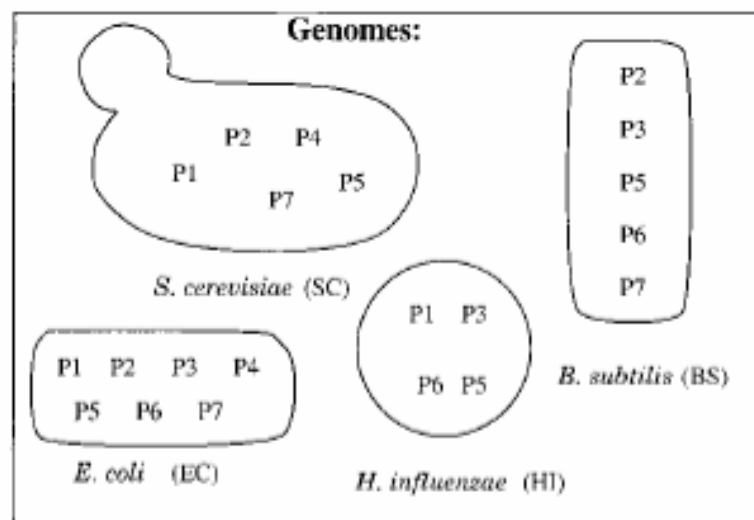
genome phylogenetic profiles
protfun's feature profiles
Similarity of dissimilarities



Phylogenetic Profiling

Pellegrini et al., *PNAS*, 96:4285--4288, 1999

- **Gene (and hence proteins) with identical patterns of occurrence across phyla tend to function together**
- ⇒ **Even if no homolog with known function is available, it is still possible to infer function of a protein**



Phylogenetic Profiling: How it Works

Phylogenetic Profiling: P-value

The probability of observing by chance z occurrences of genes X and Y in a set of N lineages, given that X occurs in x lineages and Y in y lineages is

$$P(z|N, x, y) = \frac{w_z * \overline{w}_z}{W}$$

where

No. of ways to distribute z
co-occurrences over N
lineage's

$$w_z = \binom{N}{z}$$

$$\overline{w}_z = \binom{N-z}{x-z} * \binom{N-z}{y-z}$$

No. of ways to distribute
the remaining $x - z$ and $y - z$
occurrences over the remaining
 $N - z$ lineage's

$$W = \binom{N}{x} * \binom{N}{y}$$

No. of ways of
distributing X and Y
over N lineage's
without restriction

Phylogenetic Profiles: Evidence

Pellegrini et al., *PNAS*, 96:4285--4288, 1999



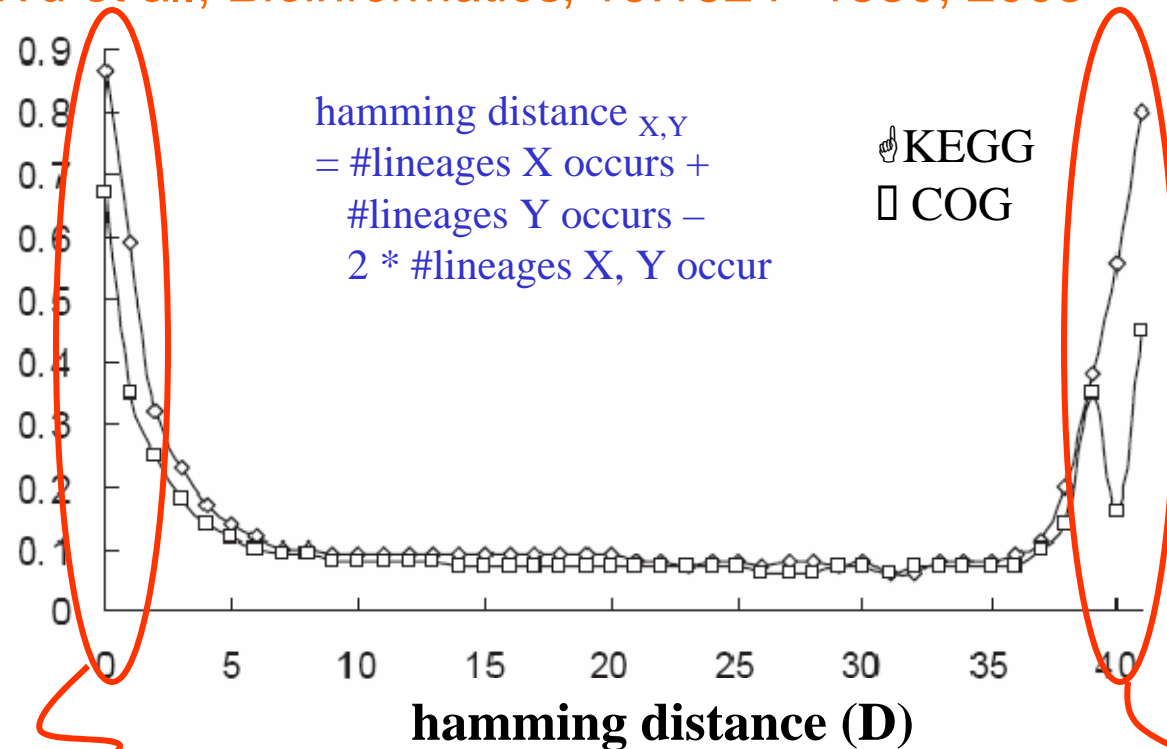
Keyword	No. of non-homologous proteins in group	No. neighbors in keyword group	No. neighbors in random group
Ribosome	60	197	27
Transcription	36	17	10
tRNA synthase and ligase	26	11	5
Membrane proteins*	25	89	5
Flagellar	21	89	3
Iron, ferric, and ferritin	19	31	2
Galactose metabolism	18	31	2
Molybdoterin and Molybdenum, and molybdoterin	12	6	1
Hypothetical†	1,084	108,226	8,440

- **E. coli proteins grouped based on similar keywords in SWISS-PROT have similar phylogenetic profiles**

Phylogenetic Profiling: Evidence

Wu et al., *Bioinformatics*, 19:1524--1530, 2003

fraction of gene pairs
having hamming distance D
and share a common pathway
in KEGG/COG



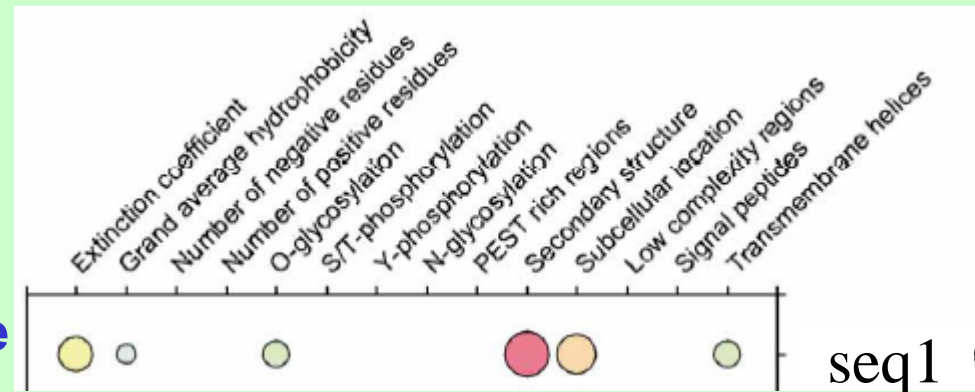
- Proteins having low hamming distance (thus highly similar phylogenetic profiles) tend to share common pathways

Exercise: Why do proteins having high hamming distance also have this behaviour?

The ProtFun Approach

Jensen, *JMB*, 319:1257--1265, 2002

- A protein is not alone when performing its biological function
- It operates using the same cellular machinery for modification and sorting as all other proteins do, such as glycosylation, phosphorylation, signal peptide cleavage, ...
- These have associated consensus motifs, patterns, etc.



- Proteins performing similar functions should share some such “features”

⇒ Perhaps we can predict protein function by comparing its “feature” profile with other proteins?

ProtFun: How it Works

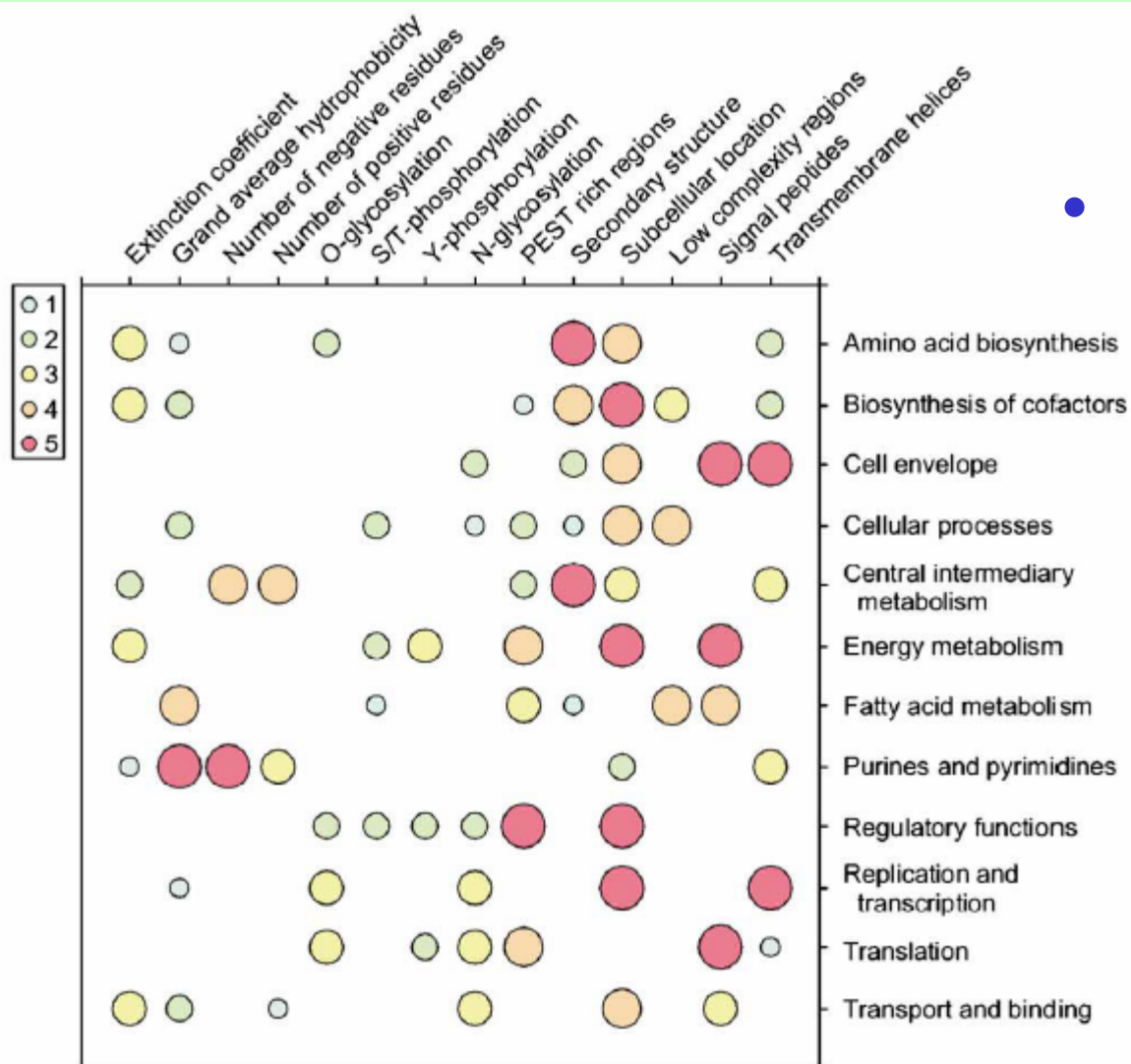
Abbreviation	Encoding	Description
ec	single value	Extinction coefficient predicted by ExPASy ProtParam
gravy	single value	Hydrophobicity predicted by ExPASy ProtParam
nneg	single value	Number of negatively charged residues counted by ExPASy ProtParam
npos	single value	Number of positively charged residues counted by ExPASy ProtParam
nglyc	potential in 5 bins	N-glycosylation sites predicted by NetNGlyc
oglyc	potential-threshold in 10 bins	GalNAc O-glycosylations predicted by NetOGlyc
pest	fraction in 10 bins	PEST rich regions identified by PESTfind
phosST	potential in 10 bins	Serine and threonine phosphorylations predicted by NetPhos
phosY	potential in 10 bins	Tyrosine phosphorylations predicted by NetPhos
psipred	helix, sheet, coil in 5 bins	Predicted secondary structure from PSI-Pred
psort	20 probabilities	Subcellular location predictions by PSORT
seg	fraction in 10 bins	Low-complexity regions identified by SEG
signalp	meanS, maxY, log(cleavage pos)	Signal peptide predictions made by SignalP
tmhmm	inside, outside, membrane in 5 bins	Transmembrane helix predictions made by TMHMM

Extract feature
profile of protein
using various
prediction methods

Category	Hidden units	Input features
Amino acid biosynthesis	30	ec psipred psort tmhmm
	30	ec psipred tmhmm
	30	ec netoglyc psipred psort
	30	gravy psipred psort
	30	oglyc psipred psort

Average the output of
the 5 component ANNs

ProtFun: Evidence



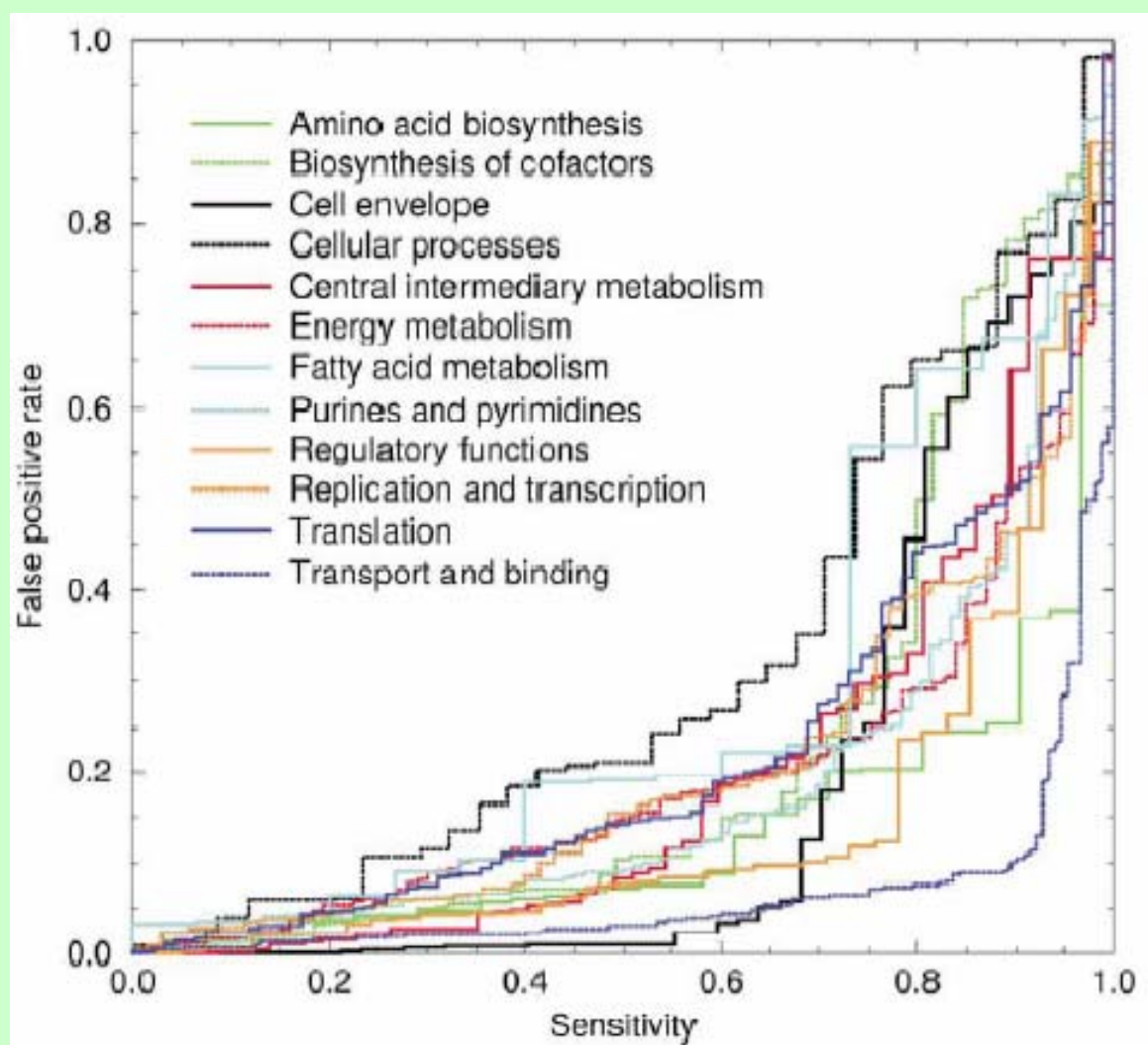
- Combinations of “features” seem to characterize some functional categories

ProtFun: Example Output

	Prion	A4	TTHY
Amino acid biosynthesis	0.011	0.011	0.011
Biosynthesis of cofactors	0.041	0.161	0.034
Cell envelope	0.146	0.804	0.698
Cellular processes	0.027	0.027	0.051
Central intermediary metabolism	0.047	0.139	0.059
Energy metabolism	0.029	0.023	0.046
Fatty acid metabolism	0.017	0.017	0.023
Purines and pyrimidines	0.528	0.417	0.153
Regulatory functions	0.013	0.014	0.014
Replication and transcription	0.020	0.029	0.040
Translation	0.035	0.027	0.032
Transport and binding	0.831	0.827	0.812
Enzyme	0.233	0.367	0.227
Non-enzyme	0.767	0.633	0.773
Oxidoreductase (EC 1.-.-.-)	0.070	0.024	0.055
Transferase (EC 2.-.-.-)	0.031	0.208	0.037
Hydrolase (EC 3.-.-.-)	0.101	0.090	0.208
Isomerase (EC 4.-.-.-)	0.020	0.020	0.020
Ligase (EC 5.-.-.-)	0.010	0.010	0.010
Lyase (EC 6.-.-.-)	0.017	0.078	0.017

- At the seq level, Prion, A4, & TTHY are dissimilar
- ProtFun predicts them to be cell envelope-related, transport & binding
- This is in agreement w/ known functionality of these proteins

ProtFun: Performance



Similarity of Dissimilarities

	orange₁	banana₁	...
apple₁	Color = red vs orange Skin = smooth vs rough Size = small vs small Shape = round vs round	Color = red vs yellow Skin = smooth vs smooth Size = small vs small Shape = round vs oblong	...
apple₂	Color = red vs orange Skin = smooth vs rough Size = small vs small Shape = round vs round	Color = red vs yellow Skin = smooth vs smooth Size = small vs small Shape = round vs oblong	...
orange₂	Color = orange vs orange Skin = rough vs rough Size = small vs small Shape = round vs round	Color = orange vs yellow Skin = rough vs smooth Size = small vs small Shape = round vs oblong	..
...

SVM-Pairwise Framework

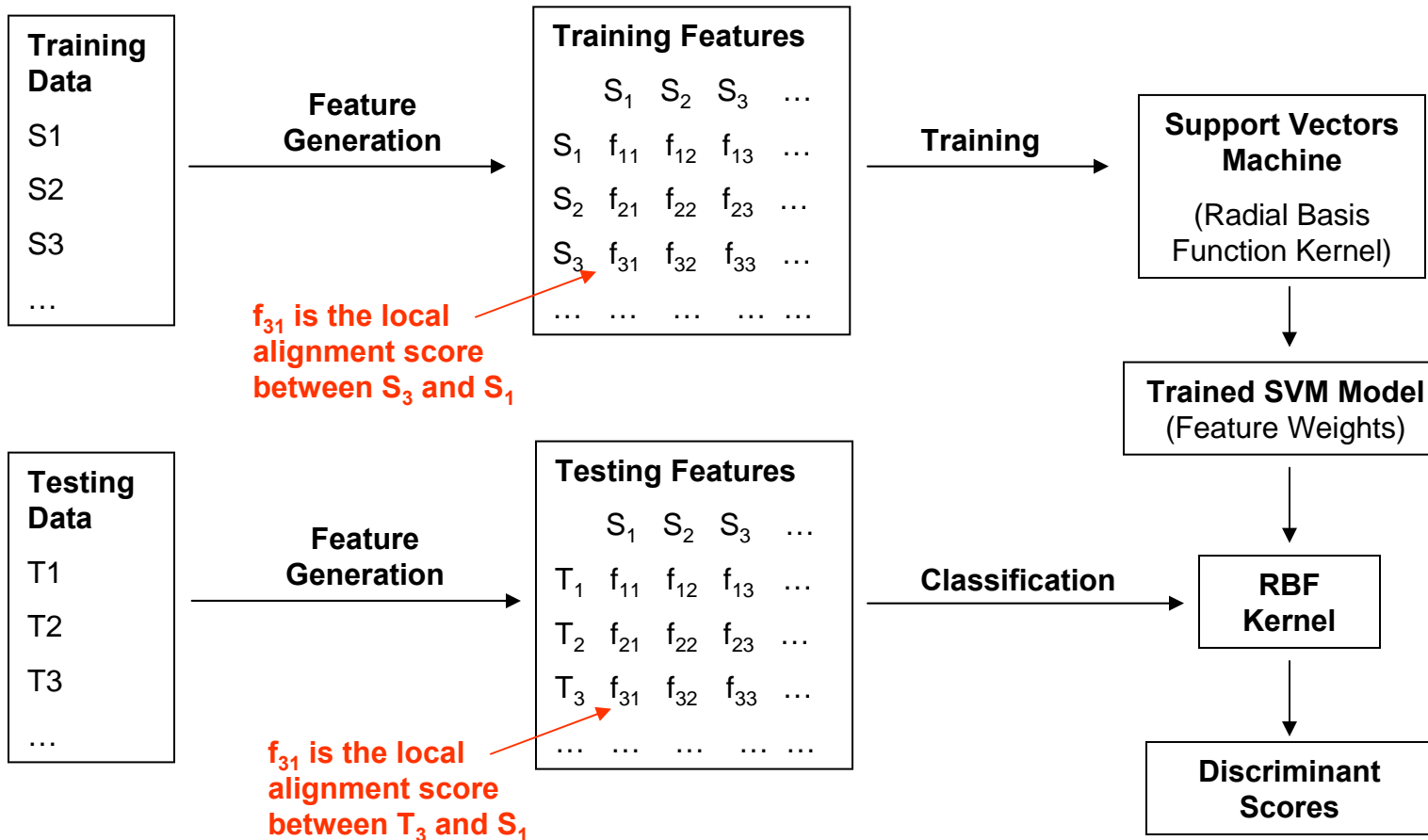
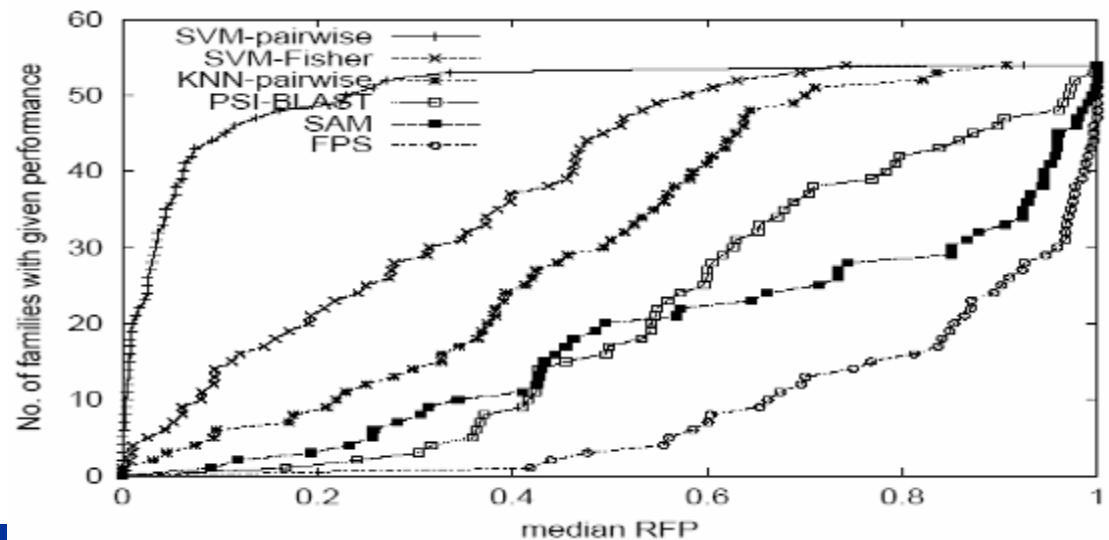
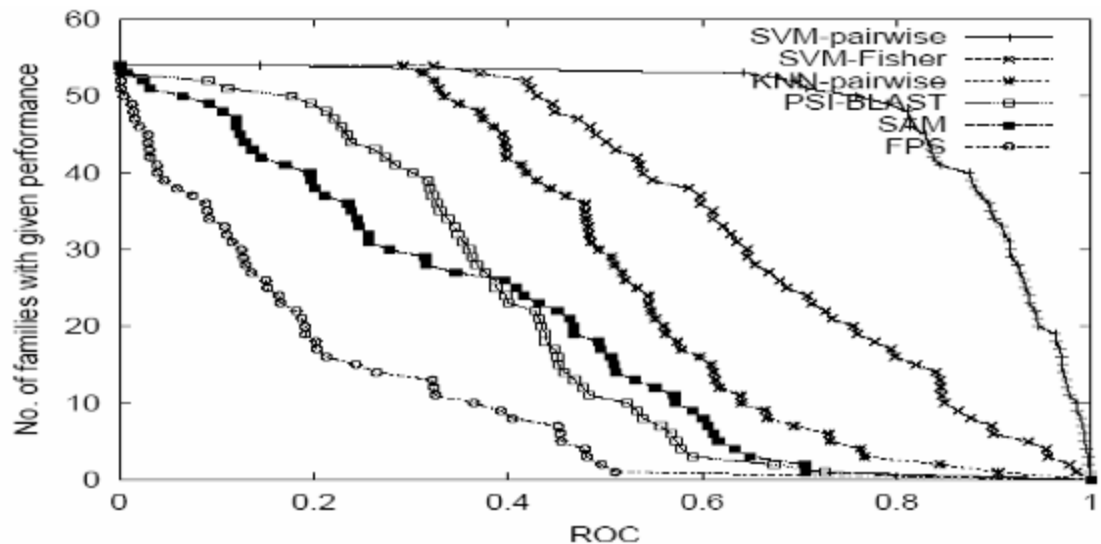


Image credit: Kenny Chua

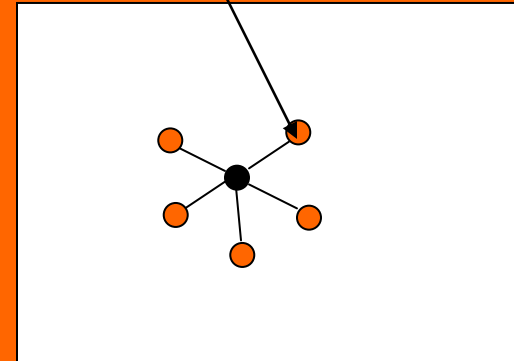
Performance of SVM-Pairwise

- **Receiver Operating Characteristic (ROC)**
 - The area under the curve derived from plotting true positives as a function of false positives for various thresholds.
- **Rate of median False Positives (RFP)**
 - The fraction of negative test examples with a score better or equals to the median of the scores of positive test examples.

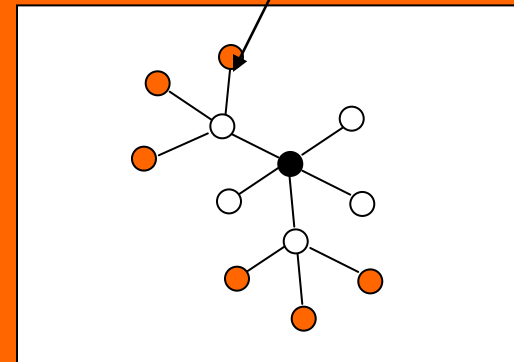


Protein Function Prediction from Protein Interactions

Level-1 neighbour



Level-2 neighbour

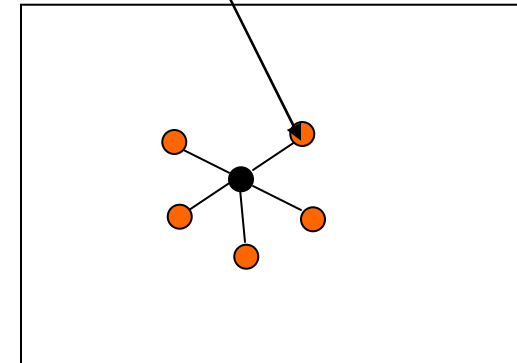


NUS
National University
of Singapore

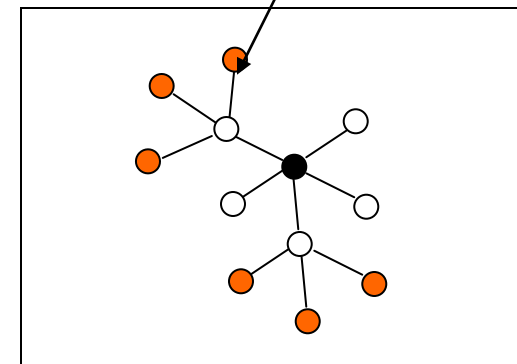
Functional Association Thru Interactions

- **Direct functional association:**
 - Interaction partners of a protein are likely to share functions w/ it
 - Proteins from the same pathways are likely to interact
- **Indirect functional association**
 - Proteins that share interaction partners with a protein may also likely to share functions w/ it
 - Proteins that have common biochemical, physical properties and/or subcellular localization are likely to bind to the same proteins

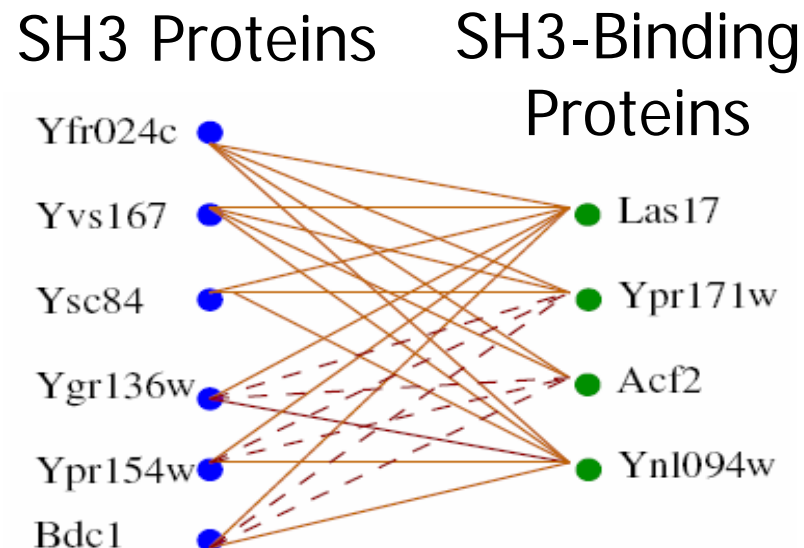
Level-1 neighbour



Level-2 neighbour

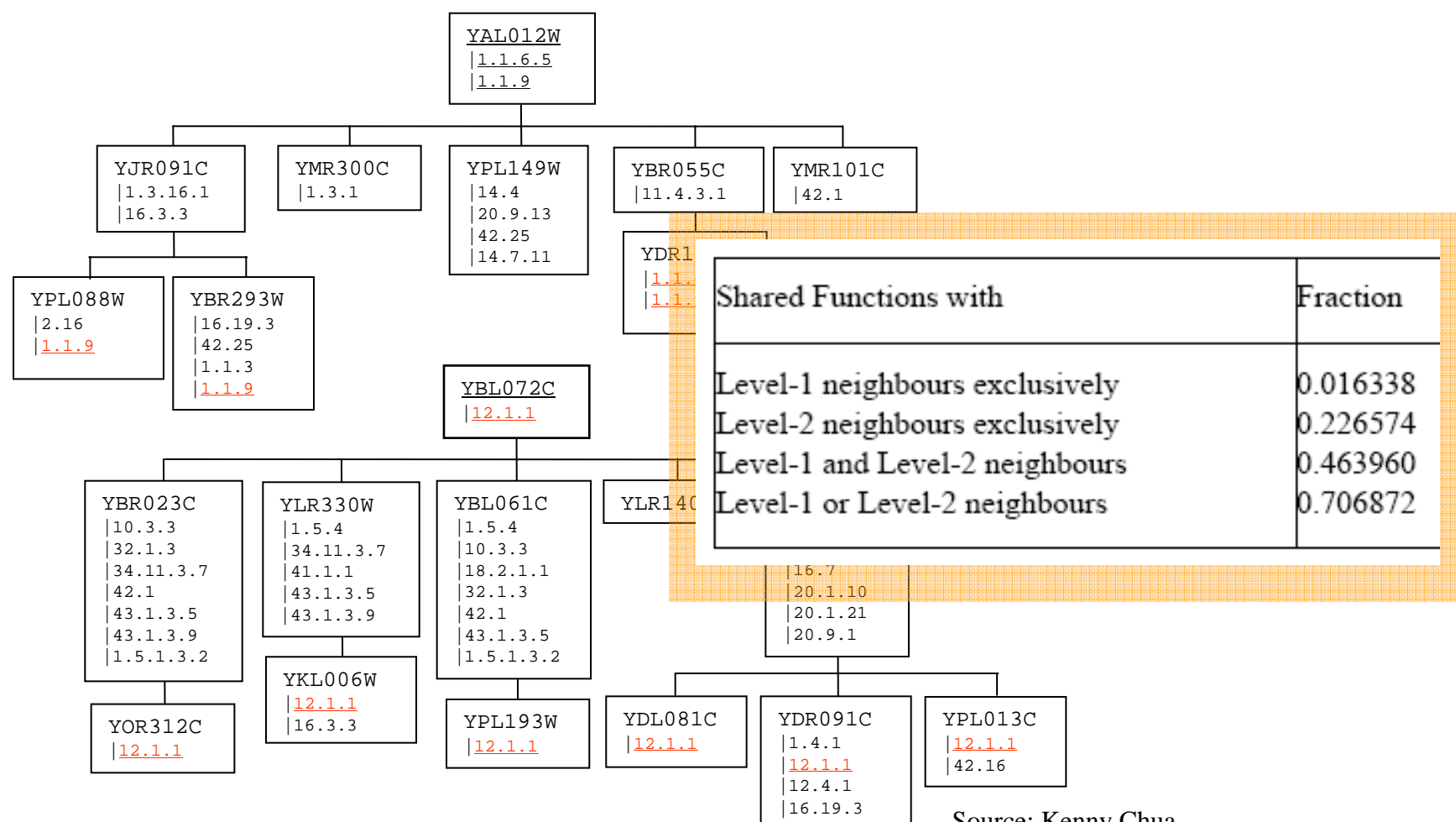


An illustrative Case of Indirect Functional Association?



- Is *indirect functional association* plausible?
- Is it found often in real interaction data?
- Can it be used to improve protein function prediction from protein interaction data?

Freq of Indirect Functional Association

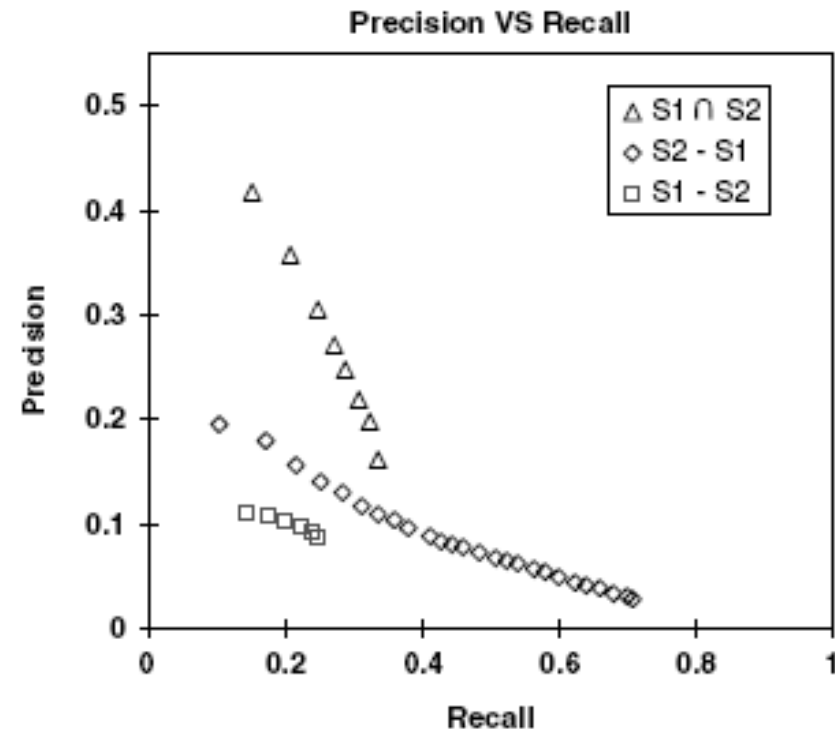


Prediction Power By Majority Voting

- Remove overlaps in level-1 and level-2 neighbours to study predictive power of “level-1 only” and “level-2 only” neighbours
- Sensitivity vs Precision analysis

$$PR = \frac{\sum_i^K k_i}{\sum_i^K m_i} \quad SN = \frac{\sum_i^K k_i}{\sum_i^K n_i}$$

- n_i is no. of fn of protein i
- m_i is no. of fn predicted for protein i
- k_i is no. of fn predicted correctly for protein i



⇒ “level-2 only” neighbours performs better

⇒ L1 ∩ L2 neighbours has greatest prediction power

Functional Similarity Estimate: Czekanowski-Dice Distance

- **Functional distance between two proteins** (Brun et al, 2003)

$$D(u, v) = \frac{|N_u \Delta N_v|}{|N_u \cup N_v| + |N_u \cap N_v|}$$

- N_k is the set of interacting partners of k
- $X \Delta Y$ is symmetric diff betw two sets X and Y
- Greater weight given to similarity

⇒ **Similarity can be defined as**

$$S(u, v) = 1 - D(u, v) = \frac{2X}{2X + (Y + Z)}$$

Is this a good measure if u and v have very diff number of neighbours?

Functional Similarity Estimate: FS-Weighted Measure



- **FS-weighted measure**

$$S(u, v) = \frac{2|N_u \cap N_v|}{|N_u - N_v| + 2|N_u \cap N_v|} \times \frac{2|N_u \cap N_v|}{|N_v - N_u| + 2|N_u \cap N_v|}$$

- N_k is the set of interacting partners of k
- Greater weight given to similarity

⇒ **Rewriting this as**

$$S(u, v) = \frac{2X}{2X + Y} \times \frac{2X}{2X + Z}$$

Correlation w/ Functional Similarity

- Correlation betw functional similarity & estimates

Neighbours	CD-Distance	FS-Weight
S_1	0.471810	0.498745
S_2	0.224705	0.298843
$S_1 \cup S_2$	0.224581	0.29629

- Equiv measure slightly better in correlation w/ similarity for L1 & L2 neighbours

Reliability of Expt Sources

- **Diff Expt Sources have diff reliabilities**
 - Assign reliability to an interaction based on its expt sources (Nabieva et al, 2004)
- **Reliability betw u and v computed by:**

$$r_{u,v} = 1 - \prod_{i \in E_{u,v}} (1 - r_i)$$

- r_i is reliability of expt source i ,
- $E_{u,v}$ is the set of expt sources in which interaction betw u and v is observed

Source	Reliability
Affinity Chromatography	0.823077
Affinity Precipitation	0.455904
Biochemical Assay	0.666667
Dosage Lethality	0.5
Purified Complex	0.891473
Reconstituted Complex	0.5
Synthetic Lethality	0.37386
Synthetic Rescue	1
Two Hybrid	0.265407

Functional Similarity Estimate: FS-Weighted Measure with Reliability

- Take reliability into consideration when computing FS-weighted measure:

$$S_R(u, v) = \frac{2 \sum_{w \in (N_u \cap N_v)} r_{u,w} r_{v,w}}{\left(\sum_{w \in N_u} r_{u,w} + \sum_{w \in (N_u \cap N_v)} r_{u,w} (1 - r_{v,w}) \right) + 2 \sum_{w \in (N_u \cap N_v)} r_{u,w} r_{v,w}} \times \frac{2 \sum_{w \in (N_u \cap N_v)} r_{u,w} r_{v,w}}{\left(\sum_{w \in N_v} r_{v,w} + \sum_{w \in (N_u \cap N_v)} r_{v,w} (1 - r_{u,w}) \right) + 2 \sum_{w \in (N_u \cap N_v)} r_{u,w} r_{v,w}}$$

- N_k is the set of interacting partners of k
- $r_{u,w}$ is reliability weight of interaction betw u and v

⇒ Rewriting

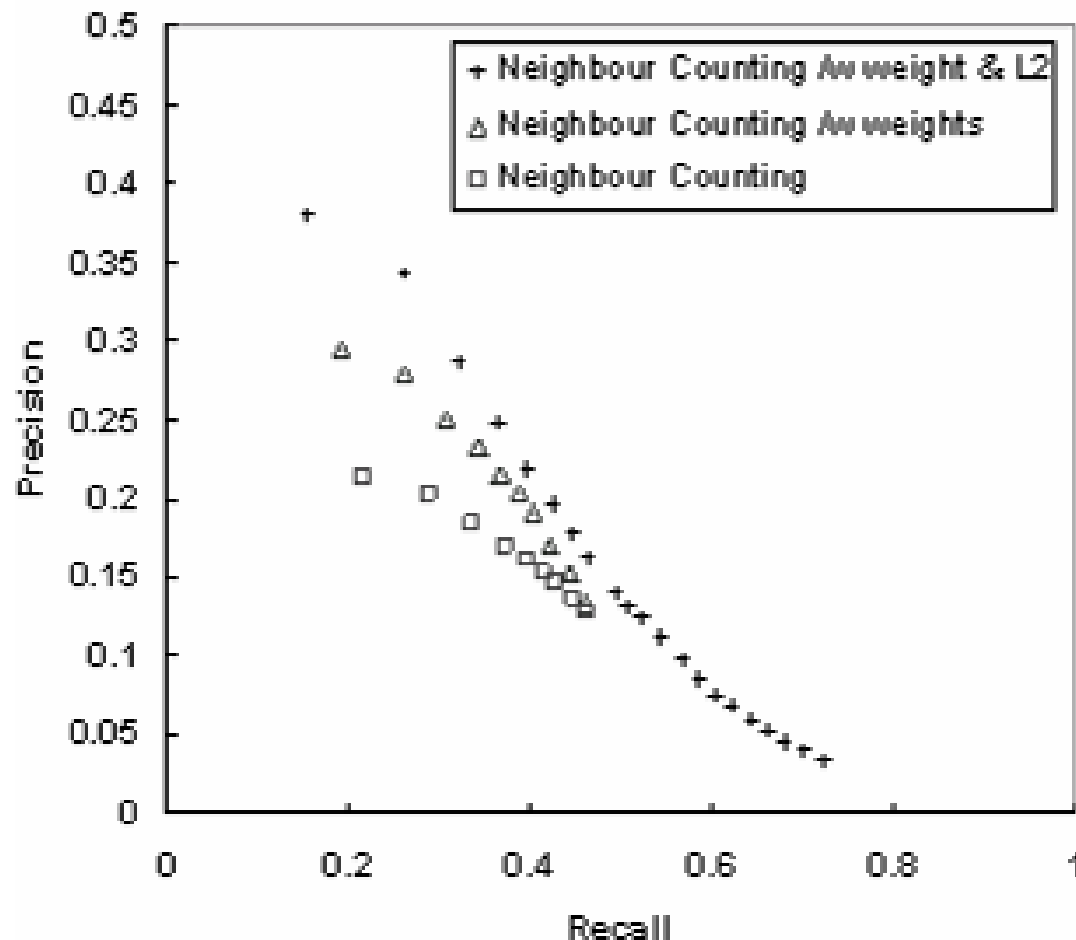
$$S(u, v) = \frac{2X}{2X + Y} \times \frac{2X}{2X + Z}$$

Integrating Reliability

- **Equiv measure shows improved correlation w/ functional similarity when reliability of interactions is considered:**

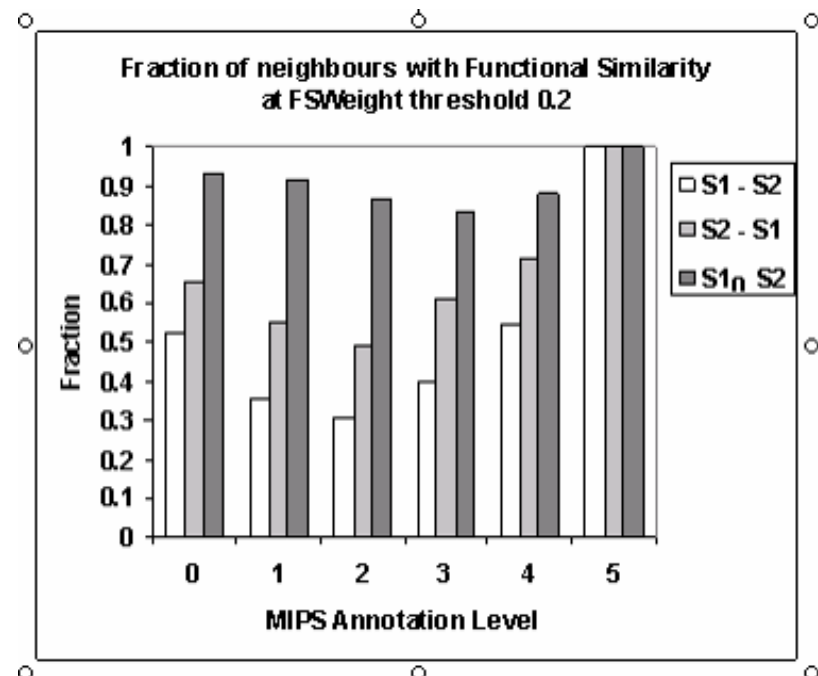
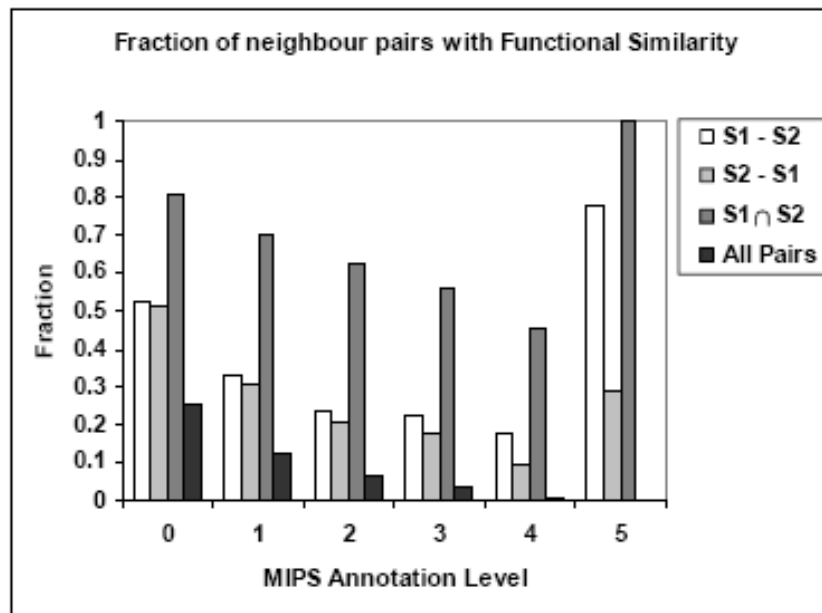
Neighbours	CD-Distance	FS-Weight	FS-Weight R
S_1	0.471810	0.498745	0.532596
S_2	0.224705	0.298843	0.375317
$S_1 \cup S_2$	0.224581	0.29629	0.363025

Improvement to Prediction Power by Majority Voting



Considering only
neighbours w/ FS
weight > 0.2

Improvement to Over-Rep of Functions in Neighbours



Use L1 & L2 Neighbours for Prediction

- **FS-weighted Average**

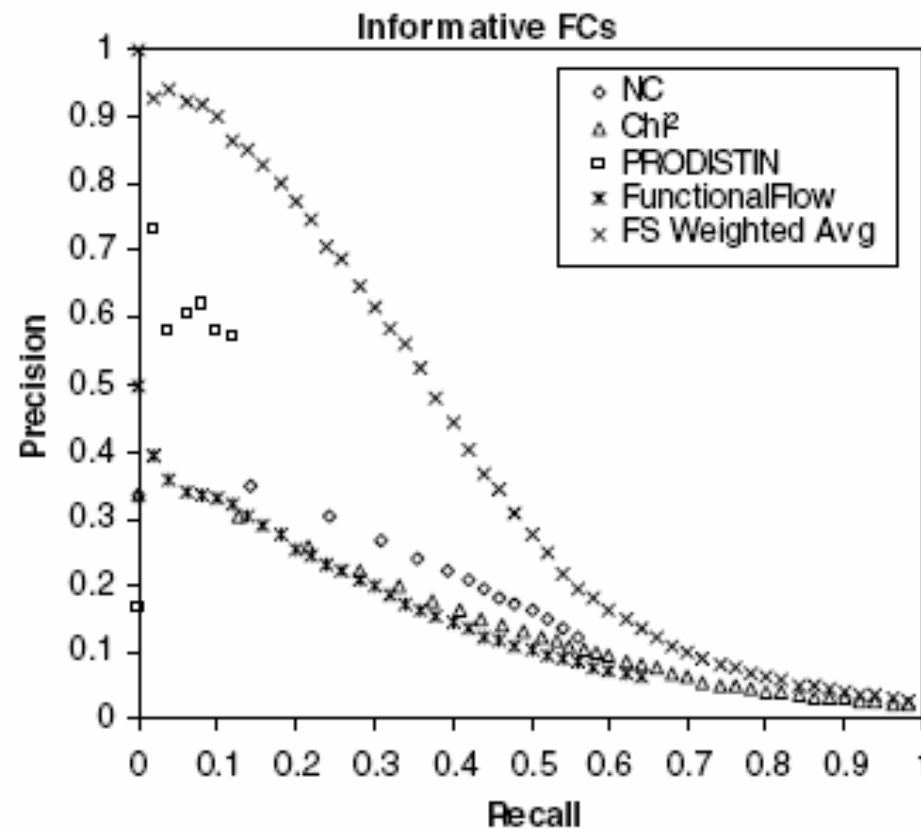
$$f_x(u) = \frac{1}{Z} \left[\lambda r_{\text{int}} \pi_x + \sum_{v \in N_u} \left(S_{TR}(u, v) \delta(v, x) + \sum_{w \in N_v} S_{TR}(u, w) \delta(w, x) \right) \right]$$

- r_{int} is fraction of all interaction pairs sharing function
- λ is weight of contribution of background freq
- $\delta(k, x) = 1$ if k has function x , 0 otherwise
- N_k is the set of interacting partners of k
- π_x is freq of function x in the dataset
- Z is sum of all weights

$$Z = 1 + \sum_{v \in N_u} \left(S_{TR}(u, v) + \sum_{w \in N_v} S_{TR}(u, w) \right)$$

Performance of FS-Weighted Averaging

- LOOCV comparison with Neighbour Counting, Chi-Square, PRODISTIN



Application of Sequence Comparison: Key Mutation Site Discovery



Identifying Key Mutation Sites

K.L.Lim et al., *JBC*, 273:28986--28993, 1998



Sequence from a typical PTP domain D2

>g1 | OCOOO | PTPA-D2

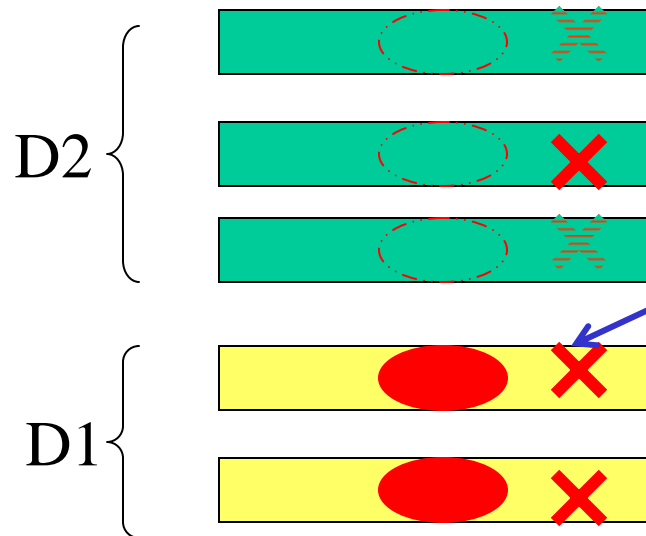
EEEFKELTSIKIQNDKERTGNLPANNEKNRVLQIIPYEFNRWIIIPVKRGEENTDYVNAIF
IDGYRQKDSYIASQGPLLHTIEDFWREIWEWKSCSIVHLTELEERGQEKCAQYWPSDGLV
SYGDIIVELKKEEECESYTVRDLLVTNNTRENKSRQIRQFHFGWPEVVGIPSDGKGNISII
LAVQKQQQQSGNEPITVHCSAGAGRTGTFCALSTVLERVKLEGILDVFQTVKSLRLQRFH
HVQTLIQYEFQYKVVQEFIDAFSDYANFK

- Some PTPs have 2 PTP domains
- PTP domain D1 is has much more activity than PTP domain D2
- Why? And how do you figure that out?

Emerging Patterns of PTP D1 vs D2

- Collect example PTP D1 sequences
- Collect example PTP D2 sequences
- Make multiple alignment A1 of PTP D1
- Make multiple alignment A2 of PTP D2
- Are there positions conserved in A1 that are violated in A2?
- These are candidate mutations that cause PTP activity to weaken
- Confirm by wet experiments

Emerging Patterns of PTP D1 vs D2



This site is consistently conserved in D1,
 but is not consistently missing in D2
 \Rightarrow it is not an EP
 \Rightarrow not a likely cause of D2's loss of function

Exercise: Why?

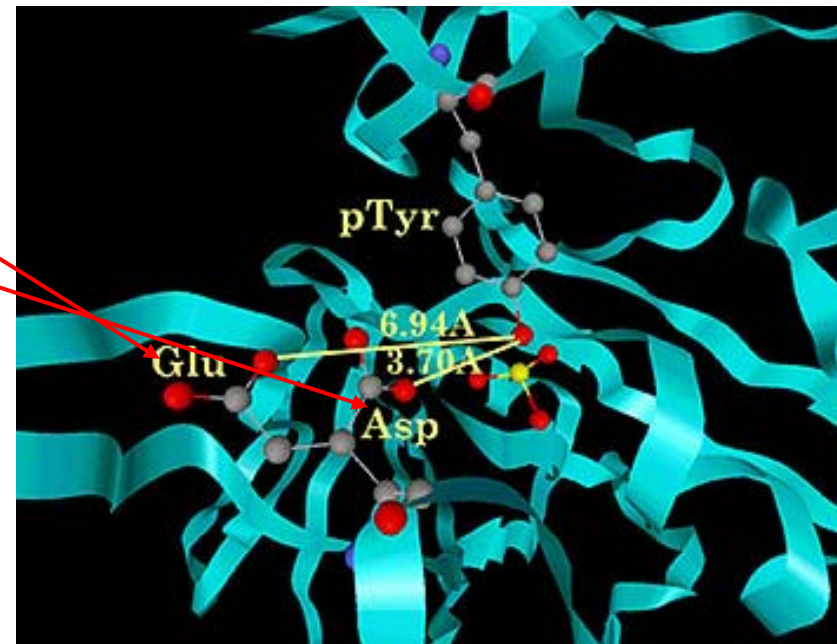
This site is consistently conserved in D1,
 but is consistently missing in D2
 \Rightarrow it is an EP
 \Rightarrow possible cause of D2's loss of function

  absent
  present

- **Positions marked by “!” and “?” are likely places responsible for reduced PTP activity**
 - All PTP D1 agree on them
 - All PTP D2 disagree on them

Key Mutation Site: PTP D1 vs D2

		?	!	?
gi 00000 P	D2	QFHFHGWPE	NG	GIPSDGK
gi 126467		Q FHFTS	W P D F	G VFFTPIC
gi 2499753		QFHFTGWP	D	HGVPHYATC
gi 462550		QYHYTQWP	D	MGVPEYALI
gi 2499751		QFHF	TSWP	DHGVPD
gi 1709906	D1	QFQFTA	WP	DHGVPEHPTI
gi 126471		QLHFTS	WP	DHGVFFTPIC
gi 548626		QFHFTGWP	DHGVPHYATC	
gi 131570		QFHFTGWP	DHGVPHYATC	
gi 2144715		QFHF	TSWP	DHGVPD
		*	..	**.*.*



- Positions marked by “!” are even more likely as 3D modeling predicts they induce large distortion to structure

Confirmation by Mutagenesis Expt

- **What wet experiments are needed to confirm the prediction?**
 - Mutate E \rightarrow D in D2 and see if there is gain in PTP activity
 - Mutate D \rightarrow E in D1 and see if there is loss in PTP activity

Exercise: Why do you need this 2-way expt?

Any Questions?



Acknowledgements

- Some of the slides are based on slides given to me by Kenny Chua

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