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

CS2220: Introduction to Computational Biology Lecture 7: Sequence Homology Interpretation

> Limsoon Wong 28 March 2008



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

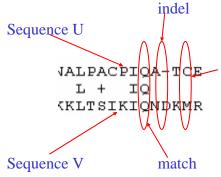
Very Brief Recap of Sequence Comparison/Alignment



Motivations for Sequence Comparis

- 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





 Key aspect of seq comparison is seq alignment

_ mismatch

 A seq alignment maximizes the number of positions that are in agreement in two sequences

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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

No obvious match between Amicyanin and Ascorbate Oxidase



- · 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]
 gi|14027493|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 MKPGRLASIALAIIFLPMAVPAHAATIEITMENLVISPTEVSAKVGDTIRWVNKDVFAHT 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
```

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Multiple Alignment: An Example



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

	/	
gi 126467	FHFTSWPDFGVPFTPIGMLKFLKKVKACNPQYAGAIVVHCSAGVGRTGTFVVIDAMLD	1
gi 2499753	FHFTGWPDHGVPYHATGLLSFIRRVKLSNPPSAGPIV <mark>V</mark> HCSA <mark>G</mark> AGRTGCYIVIDIMLD	1
gi 462550	YHYTQWPDMGVPEYALPVLTFVRRSSAARMPETGPVLVHCSAGVGRTGTYIVIDSMLQ	2
gi 2499751	FHFTSWPDHGVPDTTDLLINFRYLVRDYMKQSPPESPILVHCSAGVGRTGTFIAIDRLIY	
gi 1709906	FQFTAWPDHGVPEHPTPFLAFLRRVKTCNPPDAGPM <mark>V</mark> VHCSA <mark>G</mark> VGRTGCFIVIDAMLE	
gi 126471	LHFTSWPDFGVPFTPIGMLKFLKKVKTLNPVHAGPI <mark>V</mark> VHCSA <mark>G</mark> VGRTGTFIVIDAMMA	L
gi 548626	FHFTGWPDHGVPYHATGLLSFIRRVKLSNPPSAGPIVVHCSAGAGRTGCYIVIDIMLD	ı
gi 131570	FHFTGWPDHGVPYHATGLLGFVRQVKSKSPPNAGPLVVHCSAGAGRTGCFIVIDIMLD	ı
gi 2144715	FHFTSWPDHGVPDTTDLLINFRYLVRDYMKQSPPESPIL <mark>VHCSA</mark> GVGRTGTFIAIDRLIY	
	* *** *** . *	

Conserved sites

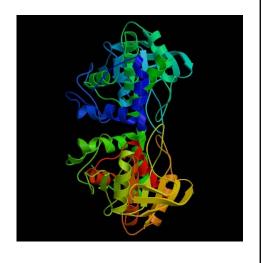
Application of Sequence Comparison: Guilt-by-Association



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
YVNILPYDHSRVHLTPVEGVPDSDYINASFINGYQEKNKFIAAQGPKEETVNDFWRMIWE
QNTATIVMVTNLKERKECKCAQYWPDQGCWTYGNVRVSVEDVTVLVDYTVRKFCIQQVGD
VTNRKPQRLITQFHFTSWPDFGVPFTPIGMLKFLKKVKACNPQYAGAIVVHCSAGVGRTG
TFVVIDAMLDMMHSERKVDVYGFVSRIRAQRCQMVQTDMQYVFIYQALLEHYLYGDTELE
VT

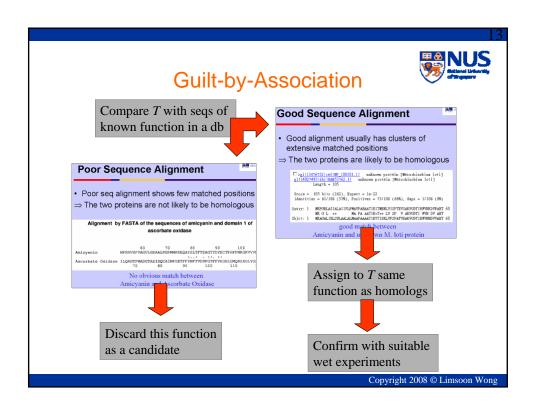
 How do we attempt to assign a function to a new protein sequence?

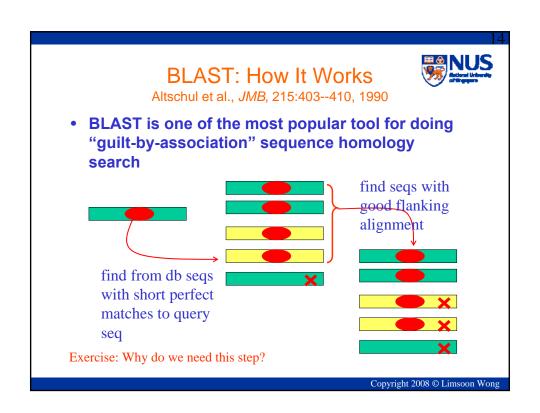
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Guilt-by-Association



- Compare the target sequence T with sequences $S_1, ..., S_n$ of known function in a database
- Determine which ones amongst S₁, ..., 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







Homologs obtained by BLAST

Sequences producing significant alignments:	(bits)	Volue
sequences producing significant alignments.	(pres)	varue
<pre>qi 14193729 qb AAK56109.1 AF332081_1 protein tyrosin phosph</pre>	62:	e-177
qi 126467 sp P18433 PTRA_HUMAN Protein-tyrosine phosphatase	<u>621 L</u>	e-177
qi 4506303 ref NP 002827.1 protein tyrosine phosphatase, r	621	e-176
gi 227294 prf 1701300A protein Tyr phosphatase	620	e-176
qi 18450369 ref NP_543030.1 protein tyrosine phosphatase,	621 L	e-176
qi 32067 emb CAA37447.1 tyrosine phosphatase precursor [Ho	61: L	e-176
<pre>qi 285113 pir JC1285 protein-tyrosine-phosphatase (EC 3.1</pre>	619	e-176
<pre>qi 6981446 ref NP_036895.1 protein tyrosine phosphatase, r</pre>	61:	e-176
gi 2098414 pdb 1YFO A Chain A, Receptor Protein Tyrosine Ph	61 S	e-174
qi 32313 emb CAA38662.1 protein-tyrosine phosphatase [Homo	61 L	e-174
<pre>qi 450583 qb AAB04150.1 protein tyrosine phosphatase >gi 4</pre>	605	e-172
qi 6679557 ref NP 033006.1 protein tyrosine phosphatase, r	<u>60. L</u>	e-172
gi 483922 gb AAA17990.1 protein tyrosine phosphatase alpha	<u>599</u>	e-170

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

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of with PTP

Example Alignment with $\text{PTP}\alpha$

```
Score = 632 bits (1629), Expect = e-180
 Identities = 294/302 (97%), Positives = 294/302 (97%)
Query: 1 SPSTNRKYPPLPVDKLEEEINRRMADDNKLFREEFNALPACP:QATCEAASXXXXXXXXXXXXX 60
           SPSTNRKVPPLPVDKLEEE INRRMADDNKLEREEEVALPACP QATCEAAS
Sbjct: 202 SPSTNRKYPPLPVDKLEEEINRRMADDNKLFREEFNALPACP:QATCEAASKEENKEKNR 261
Query: 61 YVNILPYDHSRVHLTPVEGVPDSDYINASFINGYQEKNKFIAAQGPKEETVNDFWRMIWE 120
           YVN ILPYDHSRVHLTPVEGVPDSDY I NASF I NGYQEKNKF I AAQGPKEETVNDFWRMI WE
Sbjct: 262 YVNILPYDHSRVHLTPVEGVPDSDYINASFINGYQEKNKFIAAQGPKEETVNDFWRMIWE 321
Query: 121 QNTATIVMVTNLKERKECKCAQYWPDQGCWTYGNVRVSVEDVTVLVDYTVRKFCIQQVGD 180
           QNTATIVMVTNLKERKECKCAQYWPDQGCWTYGNVRVSVEDVTVLVDYTVRKFCIQQVGD
Sbjct: 322 QNTATIVMVTNLKERKECKCAQYWPDQGCWTYGNVRVSVEDVTVLVDYTVRKFCIQQVGD 381
Query: 181 VTNRKPQRLITQFHFTSWPDFGVPFTPIGMLKFLKKVKACNPQYAGAIVVHCSAGVGRTG 240
           VTNRKPORL I TOFHFTSWPDFGVPFTP IGMLKFLKXVKACNPOYAGA I VVHCSAGVGRTG
Sbjct. 382 VTNRKPQRLITQFHFTSWPDFGVPFTPIGMLKFLKXVKACNPQYAGAIVVHCSAGVGRTG 441
Query: 241 TFVVIDAMLDMMHSERKVDVYGFVSRIRAQRCQMVQTDMQYVFIYQALLEHYLYGDTELE 300
           TFVVIDAMLD)MHSERKVDVYGFVSRIRAQRCQMVQTDMQVVFIYQALLEHYLYGDTELE
Sbjct: 442 TFVVIDAMLDMMHSERKVDVYGFVSRIRAQRCQMVQTDMQYVFIYQALLEHYLYGDTELE 501
```

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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

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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 10⁷ seqs, then you expect 10⁷ * 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

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NUS

Lightning Does Strike Twice!

- Roy Sullivan, a former park ranger from Virgina, 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



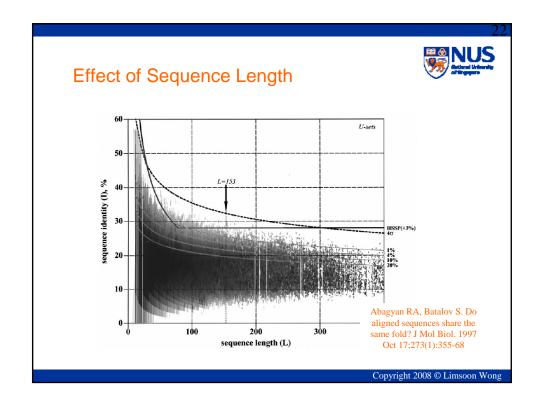
Cartoon: Ron Hipschman Data: David Hand

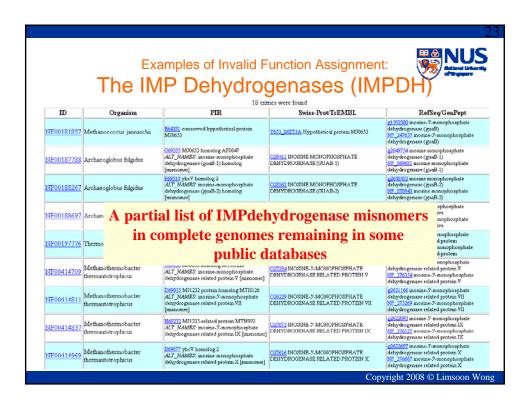
NUS

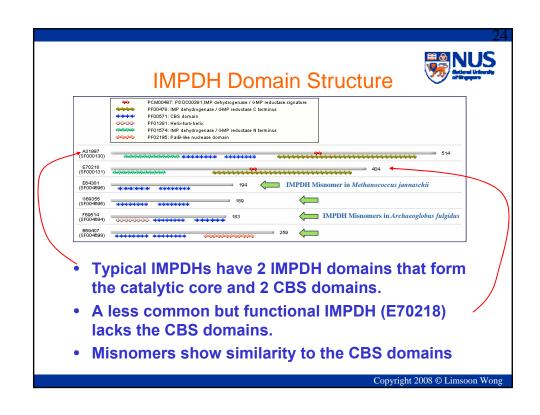
Effect of Seq Compositional Bias

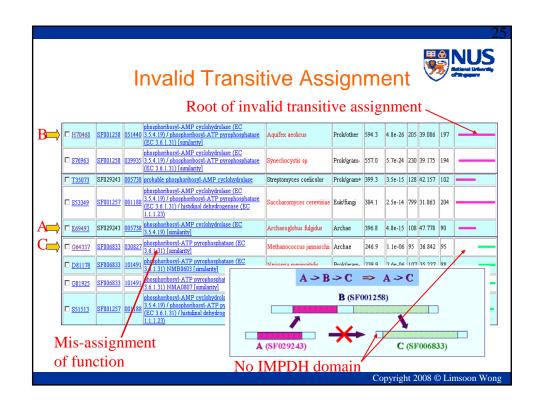
- One fourth of all residues in protein seqs occur in regions with biased amino acid composition
- Alignments of two such regions achieves high score purely due to segment composition
- ⇒ While it is worth noting that two proteins contain similar low complexity regions, they are best excluded when constructing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

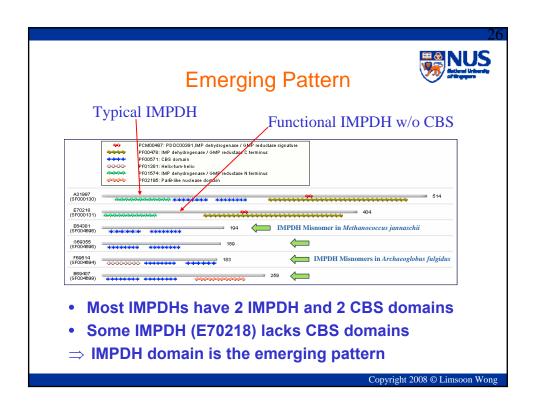
Source: NCBI











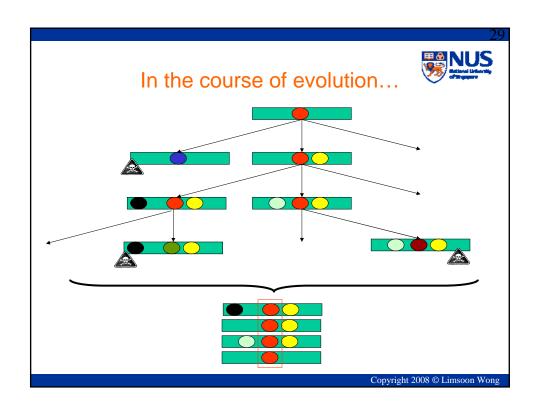
Application of Sequence Comparison: Active Site/Domain Discovery



Discover Active Site and/or Domain

- NUS
- How to discover the active site and/or domain of a function in the first place?
 - Multiple alignment of homologous seqs
 - Determine conserved positions
 - \Rightarrow 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	FHFTSWPDFGVPFTPIGMLKFLKKVKACNPQYAGAIVVHCSAGVGRTGTFVVIDAMLD
gi 2499753	FHFTGWPDHGVPYHATGLLSFIRRVKLSNPPSAGPIVVHCSAGAGRTGCYIVIDIMLD
gi 462550	YHYTQWPDMGVPEYALPVLTFVRRSSAARMPETGPVLVHCSAGVGRTGTYIVIDSMLQ
gi 2499751	FHFTSWPDHGVPDTTDLLINFRYLVRDYMKQSPPESPILVHCSAGVGRTGTFIAIDRLIY
gi 1709906	FQFTAWPDHGVPEHPTPFLAFLRRVKTCNPPDAGPMVVHCSAGVGRTGCFIVIDAMLE
gi 126471	LHFTSWPDFGVPFTPIGMLKFLKKVKTLNPVHAGPIVVHCSAGVGRTGTFIVIDAMMA
gi 548626	FHFTGWPDHGVPYHATGLLSFIRRVKLSNPPSAGPIVVHCSAGAGRTGCYIVIDIMLD
gi 131570	FHFTGWPDHGVPYHATGLLGFVRQVKSKSPPNAGPLVVHCSAGAGRTGCFIVIDIMLD
gi 2144715	${ t FHFTSWPDHGVPDTTDLLINFRYLVRDYMKQSPPESPILVHCSAGVGRTGTFIAIDRLIY}$
	* *** *** * * ***** *** **

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

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

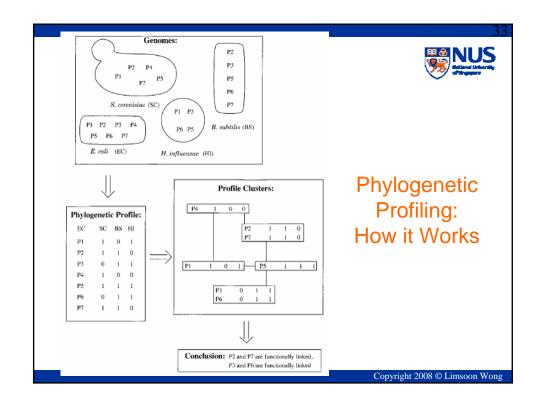
genome phylogenetic profiles protfun's feature profiles Similarity of dissimilarities



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



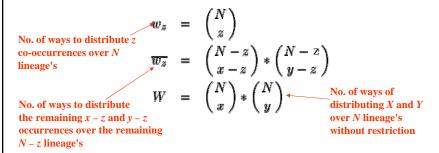
Phylogenetic Profiling: P-value



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

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

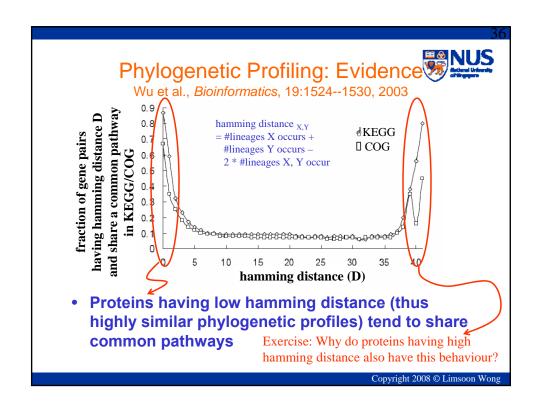
where

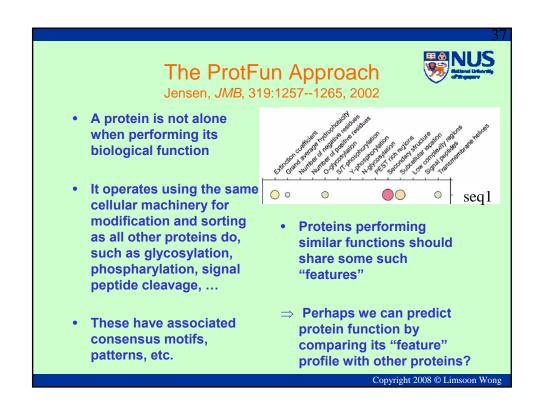


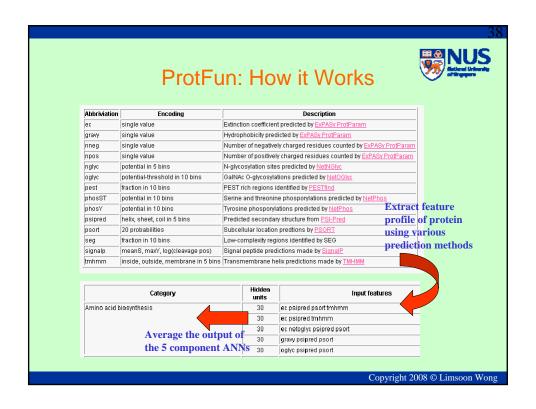
Phylogenetic Profiles: Evidence Pellegrini et al., PNAS, 96:4285--4288, 1999

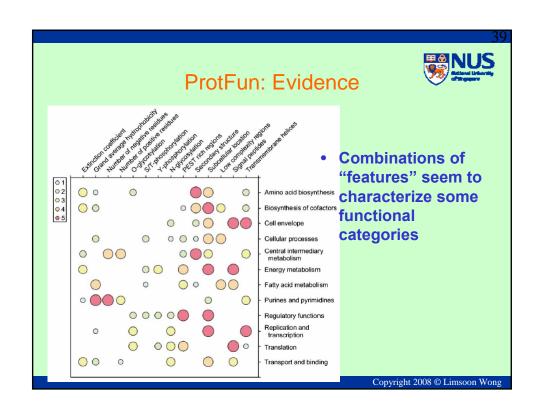
Keyword	No. of non- homologous proteins in group	No. neighbors in keyword group	No. neighbors in random group
Ribosome	69	197	27
Transcription	36	17	10
tRNA synthase and ligase	26	11	5
Membrane proteins*	25	89	5
Flagellar	21	89	3
Iron, ferric, and ferritin	19	31	2
Galactose metabolism	18	31	2
Molybdoterin and Molybdenum,			
and molybdoterin	12	6	1
Hypothetical [†]	1,084	108,226	8,440

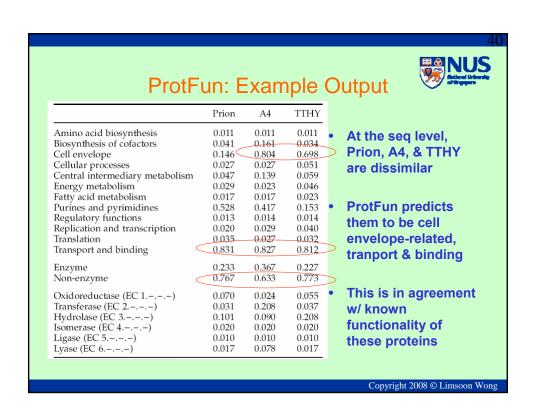
 E. coli proteins grouped based on similar keywords in SWISS-PROT have similar phylogenetic profiles

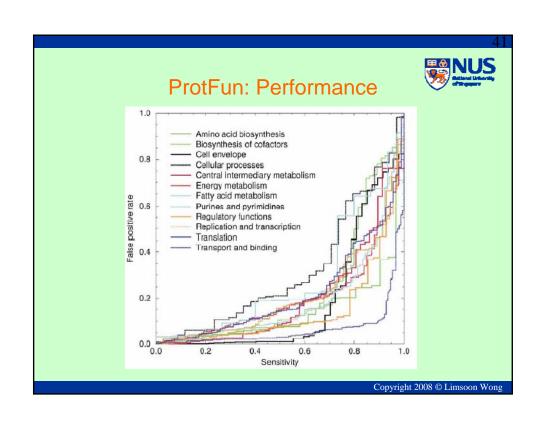




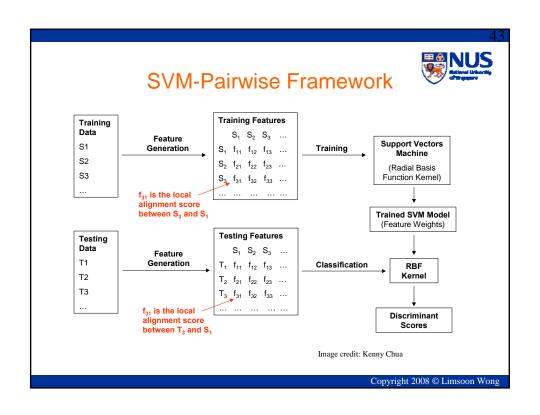


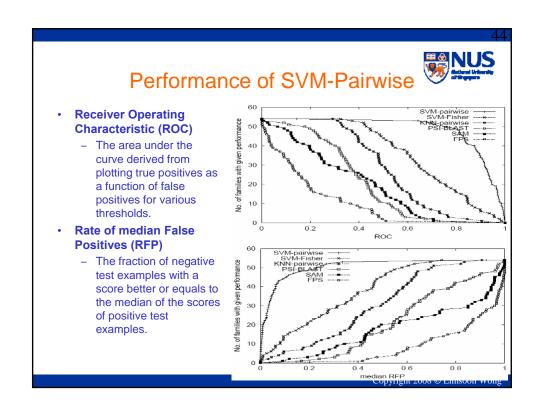


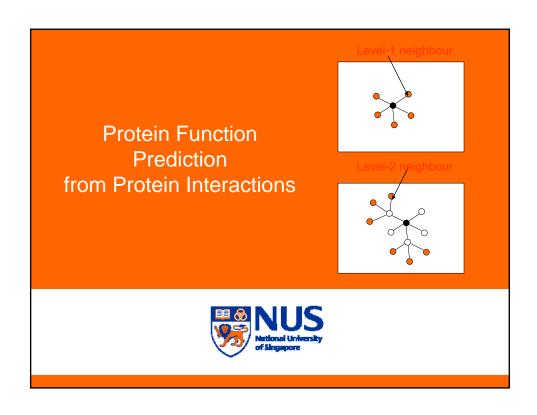


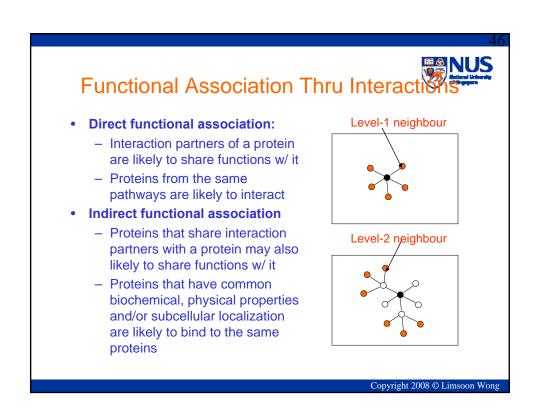


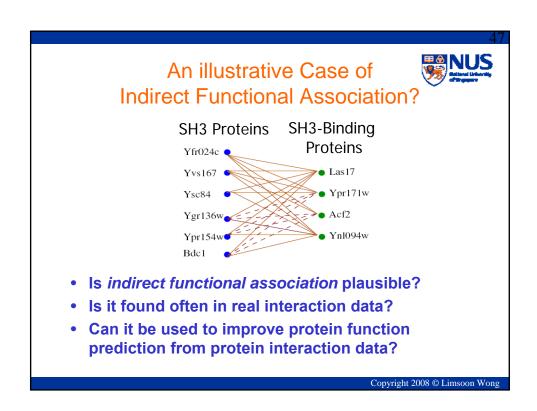
Similarity of Dissimilarities banana₁ orange₁ Color = red vs orange Color = red vs yellow apple₁ Skin = smooth vs rough Skin = smooth vs smooth Size = small vs small Size = small vs small Shape = round vs round Shape = round vs oblong Color = red vs orange Color = red vs yellow apple₂ Skin = smooth vs rough Skin = smooth vs smooth Size = small vs small Size = small vs small Shape = round vs round Shape = round vs oblong Color = orange vs orange Color = orange vs yellow orange₂ Skin = rough vs rough Skin = rough vs smooth Size = small vs small Size = small vs small Shape = round vs round Shape = round vs oblong ... Copyright 2008 © Limsoon Wong

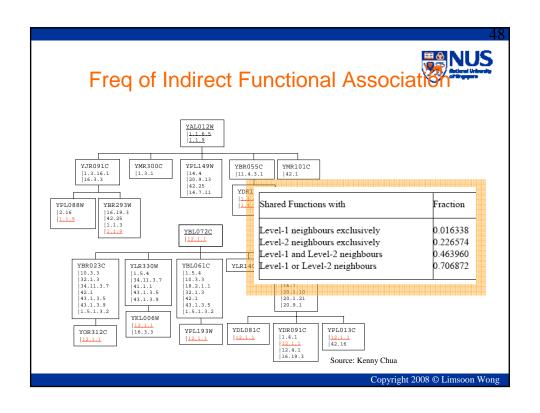










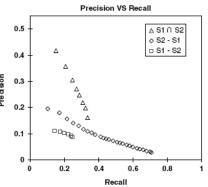




- Remove overlaps in level-1 and level-2 neighbours to study predictive power of "level-1 only" and "level-2 only" neighbours
- Sensitivity vs Precision analysis

$$PR = \frac{\sum_{i}^{K} k_{i}}{\sum_{i}^{K} m_{i}} \quad SN = \frac{\sum_{i}^{K} k_{i}}{\sum_{i}^{K} n_{i}}$$

- n; 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

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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

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

⇒ Similarity can be defined as

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

Functional Similarity Estimate: NUS 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 Similari



Correlation betw functional similarity & estimates

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

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Functional Similarity Estimate: FS-Weighted Measure with Reliability

 Take reliability into consideration when computing FS-weighted measure:

$$S_{R}(u,v) = \frac{2\sum\limits_{w \in (N_{u} \cap N_{v})} r_{u,w} r_{v,w}}{\left(\sum\limits_{w \in N_{u} - N_{v}} r_{u,w} \left(1 - r_{v,w}\right)\right) + 2\sum\limits_{w \in (N_{u} \cap N_{v})} r_{u,w}} \times \frac{2\sum\limits_{w \in (N_{u} \cap N_{v})} r_{u,w} r_{v,w}}{\left(\sum\limits_{w \in N_{v} - N_{u}} r_{v,w} \left(1 - r_{u,w}\right)\right) + 2\sum\limits_{w \in (N_{u} \cap N_{v})} r_{u,w} r_{v,w}} \times \frac{2\sum\limits_{w \in (N_{u} \cap N_{v})} r_{u,w} r_{v,w}}{\left(\sum\limits_{w \in N_{v} - N_{u}} r_{v,w} \left(1 - r_{u,w}\right)\right) + 2\sum\limits_{w \in (N_{u} \cap N_{v})} r_{u,w} r_{v,w}}$$

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

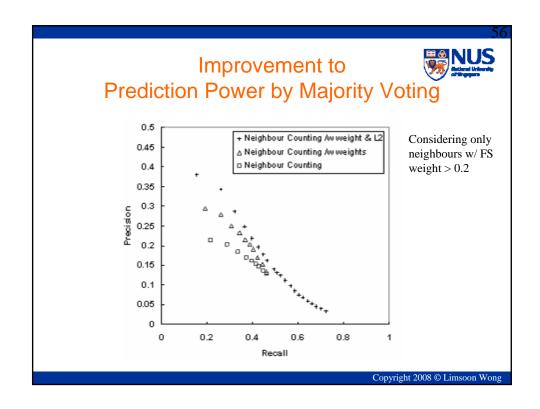
$$S(u,v) = \frac{2X}{2X+Y} \times \frac{2X}{2X+Z}$$

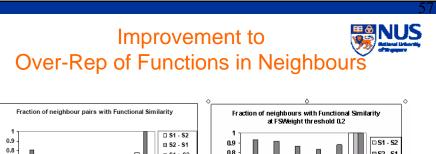


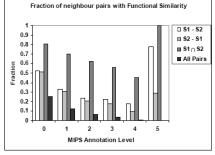
Integrating Reliability

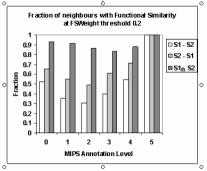
 Equiv measure shows improved correlation w/ functional similarity when reliability of interactions is considered:

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









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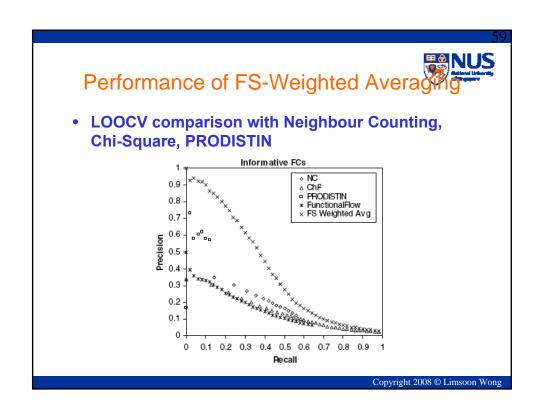
Use L1 & L2 Neighbours for Prediction

• 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,

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



Application of Sequence Comparison: Key Mutation Site Discovery





Identifying Key Mutation Sites Killim Stol. (199 STOLE) K.L.Lim et al., JBC, 273:28986--28993, 1998

Sequence from a typical PTP domain D2

>g1|00000|PTP1-02

EEEFKELTS IKIGHDKERTCHLP AMEKGNRVLQI IPYEFHRVI IPVKRGEEHTDYVMASF IDOTRORDS Y LASGOP LLETTIED FURNITURE WISCS IVELTELE EROGE RCAQTUP SOGLV Syodityelkkeeecesytyrdllythtrenksrqirqfefeoopeyoipsdokoeisii AAVQRQQQQSONEP ITVECSAGAGRTOTYCALSTVLERVKAEGILDVFQTVKSLRLQRPE EXCIT_ECTEFCYEWWQEYIDAF9DYANFK

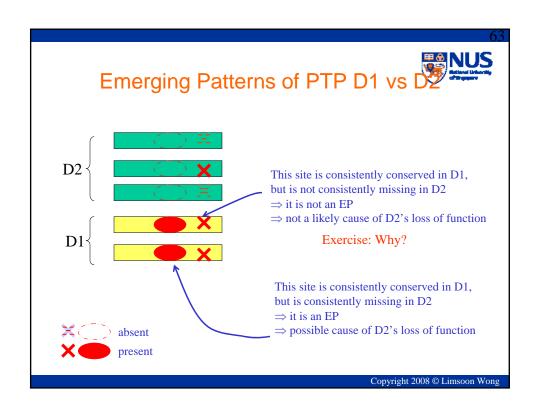
- Some PTPs have 2 PTP domains
- PTP domain D1 is has much more activity than PTP domain D2
- Why? And how do you figure that out?

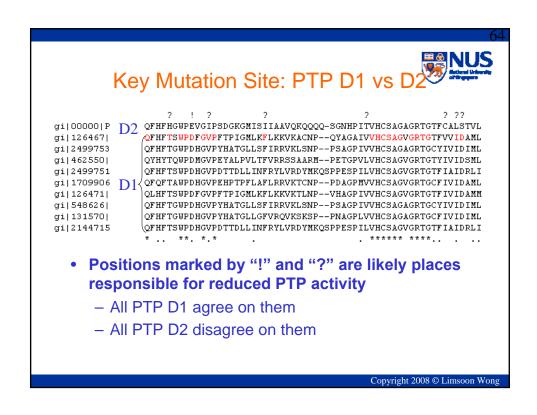
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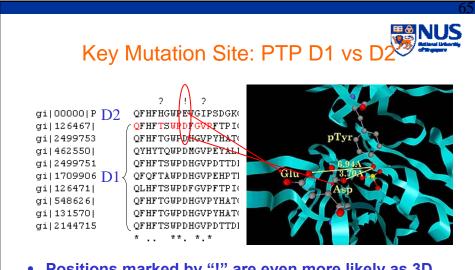
Emerging Patterns of PTP D1 vs D



- 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







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

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Confirmation by Mutagenesis Exp

- 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 → E in D1 and see if there is loss in PTP activity

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

Concluding Remarks



What have we learned?



- General methodologies & applications
 - Guilt by association for protein function inference
 - Invariants for active site discovery
 - Emerging patterns for mutation site discovery
- Important tactics
 - Genome phylogenetic profiling
 - Protfun
 - SVM-Pairwise
 - Protein-protein interactions

Any Questions?



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NUS Malana living

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