## MCB182 Class Notes: Amino Acid Similarity

<u>Similarity</u>: One of the most fundamental concepts in biology is that similarity is an indicator of common evolutionary history. Many phylogenetic trees are based on proteins. Before we can compare proteins, we need to be able to compare amino acids. How similar are any two amino acids? One could look at size, shape, charge, hydrophobicity, functional groups, etc. You can imagine a lot of different ways. The image at the right summarizes some of these ideas. Another way to determine similarity is by asking *how often can* 



one amino acid substitute for another in a protein? How would you design such an experiment? Replace an amino acid with another and perform some kind of assay for protein function? That would be great but it's a lot of work. Luckily, these experiments have already been performed billions and billions of times... in nature

<u>Margaret Dayhoff</u>: The 'mother of bioinformatics' aligned orthologous proteins by hand and published the multiple alignments in the Atlas of Protein Sequence and Structure. She and her colleagues produced multiple volumes of the atlas. Using known phylogenetic relationships, she was able to observe the rate at which one amino acid changes to another, which is called the **substitution frequency**. These changes are not symmetrical. That is, changing from G -> V  $\neq$  V -> G. This is the truth, but we generally ignore this and average them. Today, there are too many proteins for a print publication. Databases like SwissProt, GenBank, and TrEMBL take place of the Atlas. Similarly, aligning all the sequences by hand would not be possible. There are several computer programs for creating multiple alignments (ClustalW, Dialign, T-Coffee).

HM40_CAEEL/188-247	RRKRNFSKTSTEILNEYFLANINHPYPSEEVKQALAMQCNISVAQVSNWFGNKRIRYKK
PBX1_HUMAN/234-293	RRKRNFNKQATEILNEYFYSHLSNPYPSEEAKEELAKKCGITVSQVSNWFGNKRIRYKK
CUT DROME/1746-1802	KKQRVLFSEEQKEALRLAFALDPYPNVGTIEFLANELGLATRTITNWFHNHRMRLKQ
CUTL2 MOUSE/1114-1170	KKPRVVLAPAEKEALRKAYQLEPYPSQQTIELLSFQLNLKTNTVINWFHNYRSRMRR
CUTL1 MOUSE/1240-1296	KKPRVVLAPEEKEALKRAYQQKPYPSPKTIEELATQLNLKTSTVINWFHNYRSRIRR
Q22810 CAEEL/212-268	KKTKSPFTEHEIAVMMALFEINKSPNHEEVQKLAVQLNLGYRSVANFFMNKRAKERK
Q9FNM1_ARATH/51-113	PKPEWKPNQHQAQILEELFIG.GTVNPSLTSIKQITIKLQSYGEEVDDADVYKWFHNRKYSRKP
WOX9_ARATH/52-113	PKPRWNPKPEQIRILEAIFNS.GMVNPPREEIRRIRAQLQE.YGQVGDANVFYWFQNRKSRSKH
WUS_PETHY/44-106	NSTRWTPTTDQIRILKDLYYNNGVRSPTAEQIQRISAKLRQ.YGKIEGKNVFYWFQNHKARERQ
WOX6_ARATH/58-119	ATLRWNPTPEQITTLEELYRS.GTRTPTTEQIQQIASKLRK.YGRIEGKNVFYWFQNHKARERL
WOX1_ARATH/73-134	VSSRWNPTPDQLRVLEELYRQ.GTRTPSADHIQQITAQLRR.YGKIEGKNVFYWFQNHKARERQ
WOX2_ARATH/11-72	SSSRWNPTKDQITLLENLYKE.GIRTPSADQIQQITGRLRA.YGHIEGKNVFYWFQNHKARQRQ
Q8LR86_ORYSA/41-102	ANARWTPTKEQIAVLEGLYRQ.GLRTPTAEQIQQITARLRE.HGHIEGKNVFYWFQNHKARQRQ
Q9LIX7_ORYSA/24-85	STTRWCPTPEQLMMLEEMYRG.GLRTPNAAQIQQITAHLST.YGRIEGKNVFYWFQNHKARDRQ
WOX4_ARATH/87-148	GGTRWNPTQEQIGILEMLYKG.GMRTPNAQQIEHITLQLGK.YGKIEGKNVFYWFQNHKARERQ
WOX5_ARATH/21-82	KCGRWNPTVEQLKILTDLFRA.GLRTPTTDQIQKISTELSF.YGKIESKNVFYWFQNHKARERQ
WOX5_ORYSA/11-72	KCGRWNPTAEQVKVLTELFRA.GLRTPSTEQIQRISTHLSA.FGKVESKNVFYWFQNHKARERH

The **score for pairing amino acids** is shown in Equation 4. The score, *S*, for any two amino acids *i* and *j* is the log of the observed substitution frequency  $(Q_{ij})$  divided by the expected substitution frequency. The observed frequency comes from counting occurrences in multiple alignments. The expected frequency is simply the chance that any two amino acids would be selected at random, so this is the product of the probabilities of the individual amino acid frequencies  $P_i$  and  $P_j$ .

Equation 4  

$$S_{ij} = \log\left(\frac{Q_{ij}}{P_i P_j}\right)$$

 $\begin{array}{l} \textbf{Amino acid score examples} \\ \text{Given:} \quad \mathsf{P}_{M} = 0.02, \, \mathsf{P}_{L} = 0.1, \, \mathsf{P}_{E} = 0.04 \\ \quad \mathsf{Q}_{ML} = 0.004, \, \mathsf{Q}_{ME} = 0.001, \, \mathsf{Q}_{LE} = 0.002 \\ \text{Calculate:} \\ \quad S_{ML}, \, \mathsf{S}_{LE}, \, \mathsf{S}_{ME} \\ \\ \textbf{S}_{ML} = \log(0.004 / (0.02)(0.1)) = 1.0 \text{ bit} \\ \\ \textbf{S}_{LE} = \log(0.002 / (0.04)(0.1)) = -1.0 \text{ bit} \\ \\ \textbf{S}_{ME} = \log(0.001 / (0.02)(0.04)) = 0.32 \text{ bits} \\ \end{array}$ 

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Scoring matrices: A scoring matrix is simply a table of all pairwise scores. The matrix produced by Dayhoff is called the PAM matrix (a rearrangement of acceptable point mutations). If you look at the scores in a matrix, you will note that they are all integers. What happened to values like 0.32 bits? They were scaled and rounded off. For example, one might scale 0.32 by a factor of 2 and then round off 0.64 to +1. Why? Historically, computers were slow and had little memory, so people used integers. There is no reason to do this now (floating point calculations are actually faster than integer today), but the practice of using integers for scoring matrices continues. Once the scores in a matrix are scaled and rounded off, the units are no longer bits.

BLOSUM62 Scoring Matrix																		
A	R N	D	C	Q	E	G	Н	I	L	K	М	F	Р	S	Т	W	Y	۷
A 4 -	L -2	-2	0	-1	-1	0	-2	-1	-1	-1	-1	-2	-1	1	0	-3	-2	0
R -1 !	50	-2	-3	1	0	-2	0	-3	-2	2	-1	-3	-2	-1	-1	-3	-2	-3
N -2 (	56	1	-3	0	0	0	1	-3	-3	0	-2	-3	-2	1	0	-4	-2	-3
D -2 -2	2 1	6	-3	0	2	-1	-1	-3	-4	-1	-3	-3	-1	0	-1	-4	-3	-3
C 0 -2	3 - 3	-3	9	-3	-4	-3	-3	-1	-1	-3	-1	-2	-3	-1	-1	-2	-2	-1
Q -1 2	L 0	0	-3	5	2	-2	0	-3	-2	1	0	-3	-1	0	-1	-2	-1	-2
E -1 0	0 0	2	-4	2	5	-2	0	-3	-3	1	-2	-3	-1	0	-1	-3	-2	-2
G 0 -2	2 0	-1	-3	-2	-2	6	-2	-4	-4	-2	-3	-3	-2	0	-2	-2	-3	-3
H -2 0	0 1	-1	-3	0	0	-2	8	-3	-3	-1	-2	-1	-2	-1	-2	-2	2	-3
I -1 -2	3 -3	-3	-1	-3	-3	-4	-3	4	2	-3	1	0	-3	-2	-1	-3	-1	3
L -1 -2	2 -3	-4	-1	-2	-3	-4	-3	2	4	-2	2	0	-3	-2	-1	-2	-1	1
K -1 -	2 0	-1	-3	1	1	-2	-1	-3	-2	5	-1	-3	-1	0	-1	-3	-2	-2
	L - 2	-3	-1	0	-2	-3	-2	1	2	-1	5	0	-2	-1	-1	-1	-1	1
F -2 -	3 -3	-3	-2	-3	-3	-3	-1	0	0	-3	0	6	-4	-2	-2	1	3	-1
P -1 -	2 -2	-1	-3	-1	-1	-2	-2	-3	-3	-1	-2	-4	7	-1	-1	-4	-3	-2
S 1 -	L 1	0	-1	0	0	0	-1	-2	-2	0	-1	-2	-1	4	1	-3	-2	-2
T 0 -	L 0	-1	-1	-1	-1	-2	-2	-1	-1	-1	-1	-2	-1	1	5	-2	-2	0
W -3 -3	3 -4	-4	-2	-2	-3	-2	-2	-3	-2	-3	-1	1	-4	-3	-2	11	2	-3
Y -2 -3	2 -2	-3	-2	-1	-2	-3	2	-1	-1	-2	-1	3	-3	-2	-2	2	7	-1

Expected score: An important property of a matrix is its expected score (equation 5). To calculate this, one sums up the score contribution of each pairing (the contribution depends on the score and the expected frequencies of the individual amino acids). In general, the expected score of a matrix is negative.



<u>Relative entropy</u>: The most important property of a scoring matrix is its relative entropy (equation 6). This is the bits per *aligned* pair of amino acids. To gain some intuition for this, imagine if the observed pairing  $(Q_{ij})$  is equal to expected  $(P_iP_j)$ . In this case, H = 0. That is, the scoring system reflects the random expectation. This is not so different from K-L distance if you compare to identical histograms. The distance is zero. H is maximum when what is observed is very different from what is expected. When does this happen? Continuing from the previous example where  $P_M = 0.04$  and  $P_L = 0.1$ , the expectation is 0.004. If M is rarely observed to align



with L, then  $Q_{ML}$  will be different from  $P_M P_L$ . If you create a scoring matrix from proteins that are all very similar to each other, there will be few substitutions, and  $Q_{ij}$  will be very different from  $P_i P_j$ . In biological terms, a scoring matrix from highly conserved orthologous proteins will result in a matrix with high H whereas a matrix derived from less similar proteins will have low H. If the

alignments are random sequences with no real relationship, H will be zero.

<u>BLOSUM matrices</u>: Henikoff & Henikoff created their scoring matrices automatically. They did not restrict themselves to proteins with known phylogenetic relationships. To calculate the various  $Q_{ij}$  values, they assumed all pairings were possible. For any column in a multiple alignment, the counts of different amino acids is  $N_i N_j$  and the counts for the same amino acid is N choose 2. N! / 2! (N - 2)!