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Weight matrices, Sequence motifs, information content, and sequence logos

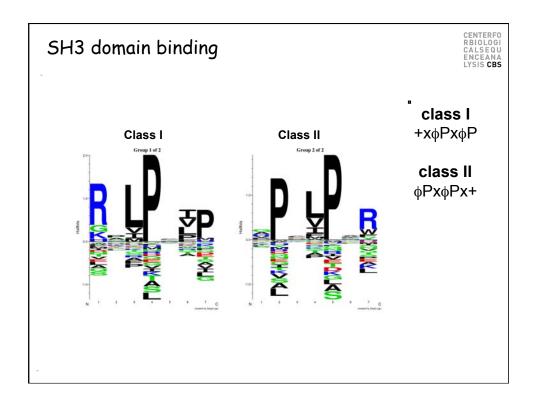
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and
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Biotecnológicas, Universidad de San Martín,
Argentina

Why weight matrices?

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- The vast majority of biological motifs are characterized by a linear motif
 - Post translational modifications
 - Signal peptides
 - T cell epitopes
 - Transcription binding sites
 - SH2/SH3 domain binding
 - MHC binding

- ...

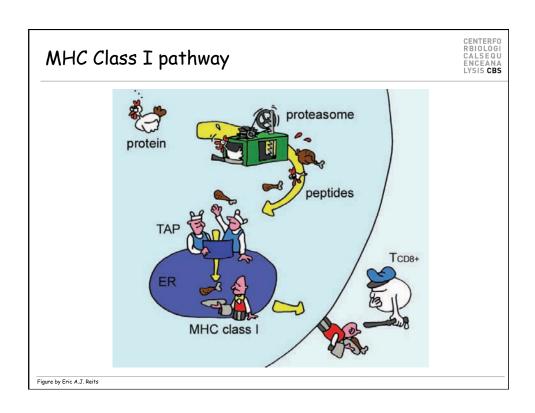


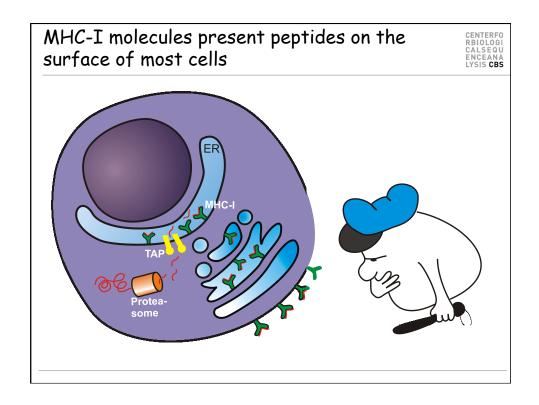
Objectives

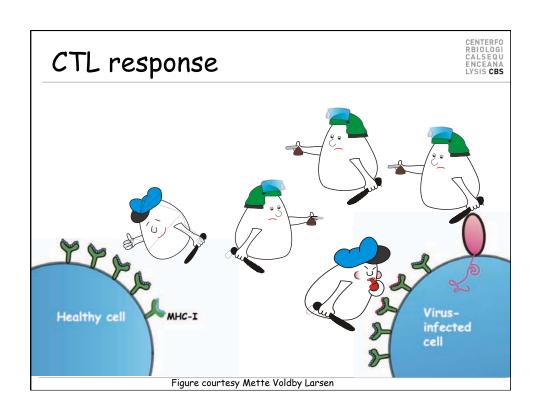
- Understand the concepts of weight matrix construction
 - One of the most important methods of bioinformatics
- · Visualization of binding motifs
 - ${\it C}{\it onstruction}$ of sequence logos
- How to construct a weight matrix
- How to use weight matrices to characterize receptor-ligand interactions
- Case story from the MHC-peptide interactions guiding immune system reactions

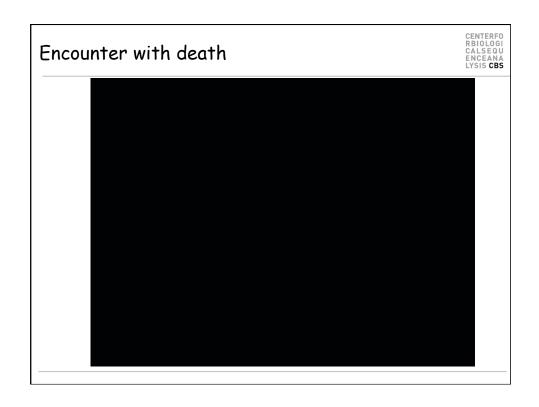
Outline

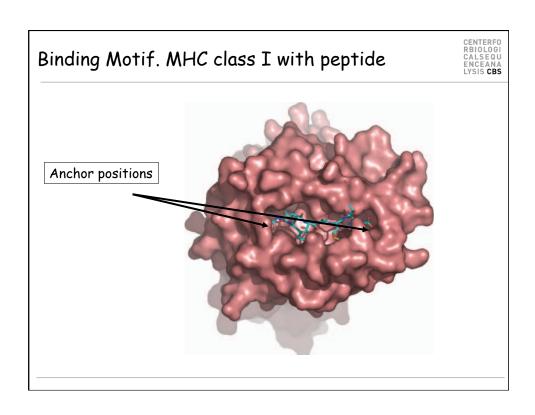
- Pattern recognition
 - Regular expressions and probabilities
- Information content
 - Sequence logos
- Multiple alignment and sequence motifs
- Weight matrix construction
 - Sequence weighting
 - Low (pseudo) counts
- Example from the real world
- · Sequence profiles
- Psi-Blast revised ...











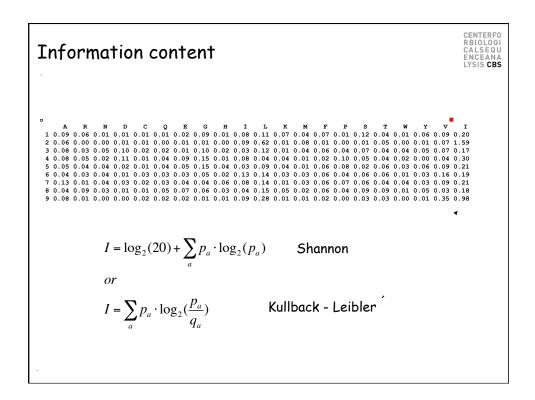
Sequence information SLLPAIVEL YLLPAIVHI TLWVDPYEV C LLDVPTAAV VLFRGGPRG MVDGTLLLL

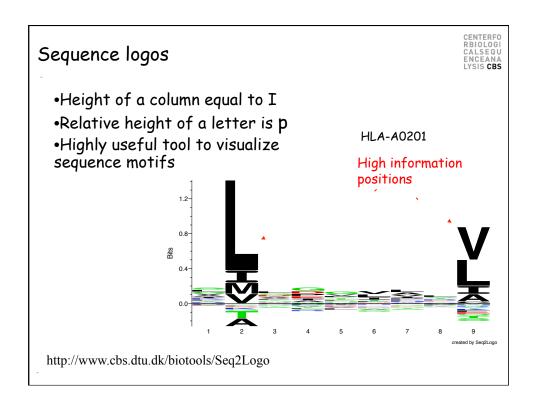
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SLLPAIVEL YLLPAIVHI TLWVDPYEV GLVPFLVSV KLLEPVILL LLDVPTAAV LLDVPTAAV LLDVPTAAV LLDVPTAAV VLFRGGRRG MVDGTLILL YMMGTMSQV MLLSVPLIL SLGGLLVEV ALLPPINIL TLIKIQHTL HLIDYLVTS ILAPPVVKL ALFPQLVIL GLGFVFTL STNRQSGRQ GLDVLTAKV RILGAVAKV QVCERTPTI ILFGGRREN ILMEHIRKI ILDQKINEV SLAGGIIGV LLIENVASL FLLWATAEA SLPDFGISY KKREEAPSL LERPGGNEI ALSNLEVKL ALNELLQHV DLERKVESL FLGENISNF ALSDHHIYL GLSEFTEYL STAPPAHGV FLOGEYFTL GVLYGVALI RTLDKVLEV HLSTAFARV RLDSYVRSL YMMGTMSQV GILGFVFTL ILKEPVHGV ILGFVFTLT LLFGYPVYV GLSPTVWLS WLSLLVPFV FLPSDFFPS CLGGLLTW FLAGNSAYE KLGEFYNQM KLVALGINA DLMGYIPLV RLVTLKDIV MLLAVLYCL AAGIGILTV YLEPGEVTA LLDGTATLR ITDQVFFSX KTWGQYWQV TITDQVPFS AFHRVAREL YLNKIQNSL MMRKLAILS ALDDKNILL IMMKNILK SMYGNNAKV SLLAPGAKQ KIFGSLĀFL ELVSEFSRM KLTPLCVTL VLYYGSFS YIGEVLVSV CINGVCWTV VMILLQYV ILTUTLGVL KVLEYVIKV FLWGPRALV GLSRYVARL FLITRILTI HLGNVKYLV GLĀGGLALL GLQDCTMLV GAGIGVAVL IAGIGILAI LIVIGIIL LAGIGLĀRĀ VDGIGIAV LVVLGLLAV ALGIGLLAV GLGCGLILQ QAGIGILLA ARGIGILAI LIVIGIIL LAGIGLĀRĀ VDGIGITU REPGRĀFV GLBCGLEVQLV PLKQHFQUL VLKQHFQUL VLKGHFQUL VLKG

Cost of a motif characterization

- 200 peptides needed
 - -50-200 \$ per peptide = 10,000 40,000 \$
 - 1 PhD student manpower
- 2000 MHC class I molecules
 - So do the math your self ...





Sequence Information

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- Say that a peptide must have L at P_2 in order to bind, and that A,F,W,and Y are found at P_1 . Which position has most information?
- How many questions do I need to ask to tell if a peptide binds looking at only P_1 or P_2 ?

Sequence Information

- Say that a peptide must have L at P₂ in order to bind, and that A,F,W,and Y are found at P₁. Which position has most information?
- How many questions do I need to ask to tell if a peptide binds looking at only P_1 or P_2 ?
- P1: 4 questions (at most)
- P2: 1 question (L or not)
- P2 has the most information

Sequence Information

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- Say that a peptide must have L at P₂ in order to bind, and that A,F,W,and Y are found at P₁. Which position has most information?
- How many questions do I need to ask to tell if a peptide binds looking at only P₁ or P₂?
- P1: 4 questions (at most)
- P2: 1 question (L or not)
- P2 has the most information

- Calculate p_a at each position
- Entropy

$$S = -\sum_{a} p_a \log(p_a)$$

Information content

$$I = \log(20) + \sum_a p_a \cdot \log(p_a)$$

or

$$I = \sum_{a} p_a \cdot \log(\frac{p_a}{q_a})$$

- Conserved positions
 - P_L=1, P_{IL}=0 => S=0, I=log(20)
- · Mutable positions
 - $P_{aa}=1/20 \Rightarrow S=log(20), I=0$

Characterizing a binding motif from small data sets

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10 MHC restricted peptides

ALAKAAAAM
ALAKAAAAAT
ALAKAAAAAT
ALAKAAAAV
GMNERPILT
GILGFVFTM
TLNAWVKVV
KLNEPVLLL
AVVPFIVSV

What can we learn?

- 1. A at P1 favors binding?
- 2. I is not allowed at P9?
- 3. K at P4 favors binding?
- 4. Which positions are important for binding?

Simple motifs

Yes/No rules

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10 MHC restricted peptides

ALAKAAAAM ALAKAAAAN ALAKAAAAR ALAKAAAAT ALAKAAAAV GMNERPILT GILGFVFTM TLNAWVKVV KLNEPVLLL **AVVPFIVSV** $[AGTK]_1[LMIV]_2[ANLV]_3...[MNRTVL]_9$

- Only 11 of 212 peptides identified!
- · Need more flexible rules •If not fit P1 but fit P2 then ok
- · Not all positions are equally important ·We know that P2 and P9 determines binding more than other positions
- ·Cannot discriminate between good and very good binders

Simple motifs

Yes/No rules

10 MHC restricted peptides

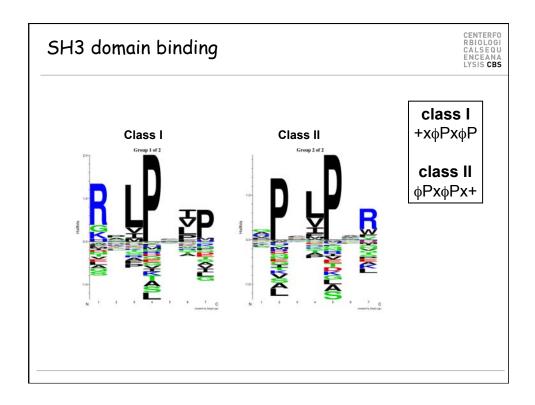
ALAKAAAAM ALAKAAAAN ALAKAAAAR ALAKAAAAT ALAKAAAAV GMNERPILT **GILGFVFTM** TLNAWVKVV KLNEPVLLL **AVVPFIVSV**

 $[AGTK]_1[LMIV]_2[ANLV]_3...[AIFKLV]_7...[MNRTVL]_9$

Example

RLLDDTPEV 84 nM GLLGNVSTV 23 nM ALAKAAAAL 309 nM

•Two first peptides will not fit the motif. They are all good binders (aff< 500nM)



Extended motifs

- Fitness of aa at each position given by P(aa)
- Example P1

$$P_A = 6/10$$

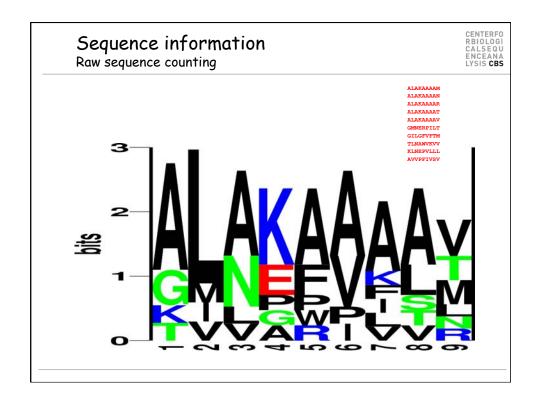
$$P_G = 2/10$$

 $P_T = P_K = 1/10$
 $P_C = P_D = ...P_V = 0$

- Problems
 - Few data
 - Data redundancy/duplication

RLLDDTPEV 84 nM GLLGNVSTV 23 nM

ALAKAAAAL 309 nM



Sequence weighting

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- •Poor or biased sampling of sequence space
- •Example P1

$$P_A = 2/6$$

$$P_G = 2/6$$

$$P_T = P_K = 1/6$$

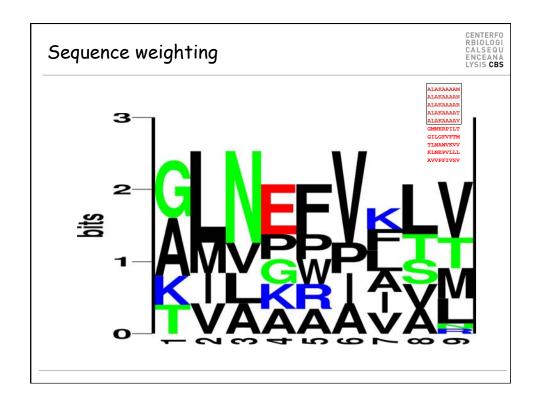
$$P_C = P_D = ...P_V = 0$$

ALAKAAAAM
ALAKAAAAAT
ALAKAAAAAT
ALAKAAAAAV
GMNERPILT
GILGFVFTM
TLNAWVKVV
KLNEPVLLL

AVVPFIVSV

Similar sequences Weight 1/5

RLLDDTPEV 84 nM GLLGNVSTV 23 nM ALAKAAAAL 309 nM



Pseudo counts

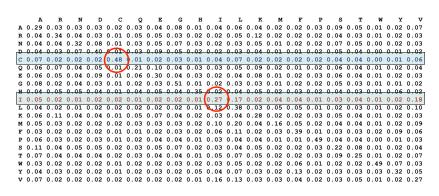
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•I is not found at position P9. Does this mean that I is forbidden (P(I)=0)?
•No! Use Blosum substitution matrix to estimate pseudo frequency of I at P9

ALAKAAAAN
ALAKAAAAR
ALAKAAAAT
ALAKAAAAV
GMNERPILT
GILGFVFTM
TLNAWVKVV
KLNEPVLIL
AVVPFIVSV

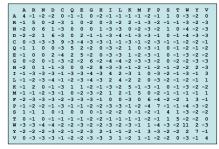
The Blosum matrix

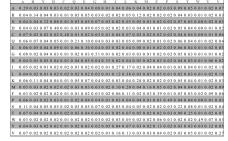
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Some amino acids are highly conserved (i.e. \mathcal{C}), some have a high change of mutation (i.e. \mathcal{I})

The way from log-odds to frequencies





$$S_{ij} = 2 \cdot \log_2(\frac{P_{ij}}{Q_i \cdot Q_j}) = 2 \cdot \log_2(\frac{P(j \mid i)}{Q_j})$$

$$S_{AA} = 2 \cdot \log_2(\frac{P(A \mid A)}{Q_A}) = 2 \cdot \log_2(\frac{0.29}{0.074}) = 3.9$$

$$S_{AR} = 2 \cdot \log_2(\frac{P(R \mid A)}{Q_A}) = 2 \cdot \log_2(\frac{0.03}{0.052}) = -1.6$$

What is a pseudo count?

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- Say V is observed at P2
- Knowing that V at P2 binds, what is the probability that a peptide could have I at P2?
- P(I|V) = 0.16

Pseudo count estimation

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- Calculate observed amino acids frequencies f_a
- Pseudo frequency for amino acid b

ALAKAAAAN
ALAKAAAAAT
ALAKAAAAAT
ALAKAAAAT
GMNERPILT
GILGFVFTM
TLNAWVKVV
KLNEPVLLL
AVVPFIVSV

$$g_b = \sum_a f_a \cdot q_{b|a}$$

• Example

$$g_I = 0.2 \cdot q_{I|M} + 0.1 \cdot q_{I|R} + ... + 0.3 \cdot q_{I|V} + 0.1 \cdot q_{I|L}$$

 $g_I = 0.2 \cdot 0.1 + 0.1 \cdot 0.02 + ... + 0.3 \cdot 0.16 + 0.1 \cdot 0.12 = 0.094$

Weight on pseudo count

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- Pseudo counts are important when only limited data is available
- ALAKAAAAN
 ALAKAAAAR
 ALAKAAAAT
 ALAKAAAAV
 GMNERPILT
 GILGFVFTM
 TLNAWVKVV
 KLNEPVLLL
- With large data sets only ☐ rue☐ observation should count

$$p_a = \frac{\alpha \cdot f_a + \beta \cdot g_a}{\alpha + \beta}$$

• α is the effective number of sequences (N-1), β is the <u>weight on prior</u> or <u>weight on pseudo count</u>

Weight on pseudo count

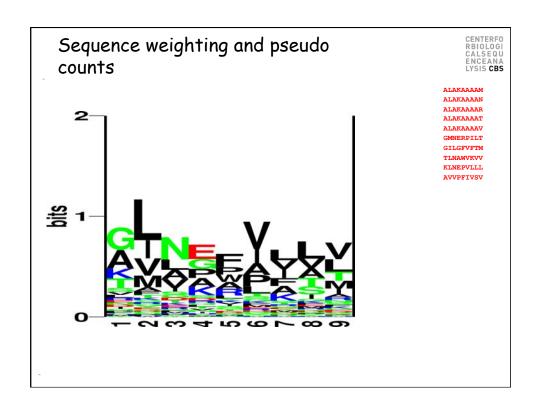
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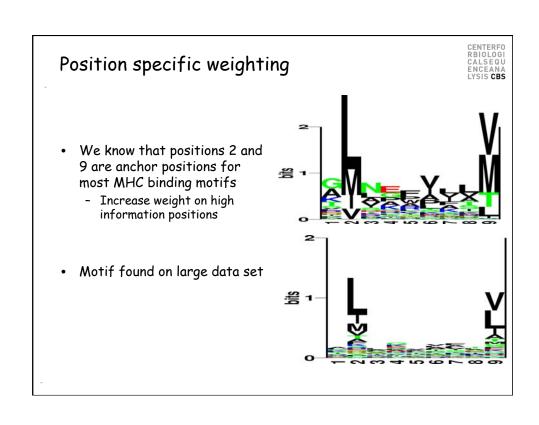
Example

$$p_a = \frac{\alpha \cdot f_a + \beta \cdot g_a}{\alpha + \beta}$$

ALAKAAAAM
ALAKAAAAA
ALAKAAAAT
ALAKAAAAT
GNNERPILT
GILGFVFTW
TLNAWVKVV
KLNEPVLLL
AVVPFIVSV

- If α large, p \approx f and only the observed data defines the motif
- If α small, $p \approx g$ and the pseudo counts (or prior) defines the motif
- β is [50-200] normally





Weight matrices

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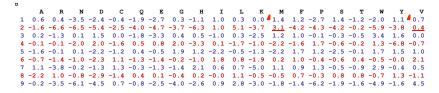
• Estimate amino acid frequencies from alignment including sequence weighting and pseudo count

- What do the numbers mean?
 - P2(V)>P2(M). Does this mean that V enables binding more than M.
 - In nature not all amino acids are found equally often
 - In nature V is found more often than M, so we must somehow rescale with the background
 - $q_M = 0.025, q_V = 0.073$
 - Finding 7% V is hence not significant, but 7% M highly significant

Weight matrices

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- · A weight matrix is given as
 - $W_{ij} = log(p_{ij}/q_j)$
 - where i is a position in the motif, and j an amino acid. q_j is the background frequency for amino acid j.



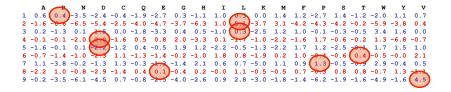
• W is a L x 20 matrix, L is motif length

18

Scoring a sequence to a weight matrix

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 Score sequences to weight matrix by looking up and adding L values from the matrix



RLLDDTPEV

11.9 84nM

GLLGNVSTV 14.7 23nM

ALAKAAAAL 4.3 309nM

Which peptide is most likely to bind? Which peptide second?

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An example!! (See handout)

Estimation of pseudo counts

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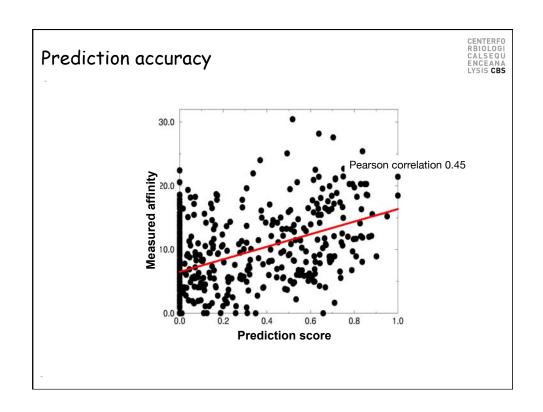
				•
	f_a	g_a	p_a	w_a
A	0	0.06	0.03	-2.61
R	0	0.053	0.027	-1.93
N	0	0.04	0.02	-2.33
D	0	0.083	0.042	-0.75
C	0	0.01	0.005	-4.64
Q	0.167	0.085	0.126	3.78
E	0.833	0.267	0.550	6.70
G	0	0.04	0.02	-3.78
Н	0	0.03	0.015	-1.59
I	0	0.022	0.011	-5.30
L	0	0.042	0.021	-4.50
K	0	0.082	0.041	-1.01
M	0	0.012	0.006	-4.19
F	0	0.018	0.009	-4.72
P	0	0.028	0.014	-2.92
S	0	0.06	0.03	-1.85
T	0	0.04	0.02	-2.70
W	0	0.01	0.005	-2.76
Y	0	0.02	0.01	-3.36
V	0	0.032	0.016	-4.41

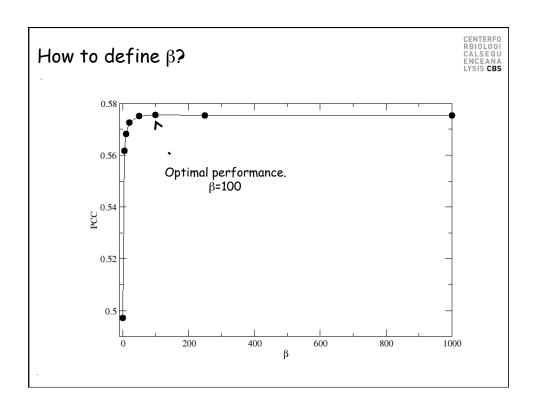
Example from real life

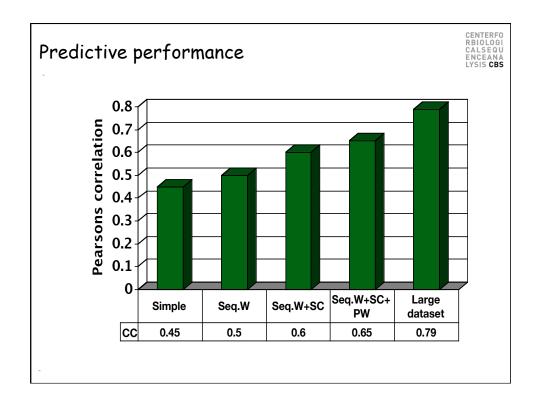
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- 10 peptides from MHCpep database
- Bind to the MHC complex
- Relevant for immune system recognition
- Estimate sequence motif and weight matrix
- Evaluate motif "correctness" on 528 peptides

ALAKAAAAM
ALAKAAAAAT
ALAKAAAAT
ALAKAAAAV
GMNERPILT
GILGFVFTM
TLNAWVKVV
KLNEPVLLL
AVVPFIVSV







Summary

- Sequence logo is a power tool to visualize (binding) motifs
 - Information content identifies essential residues for function and/or structural stability
- Weight matrices can be derived from very limited number of data using the techniques of
 - Sequence weighting
 - Pseudo counts

Sequence Profiles and Weight matrices

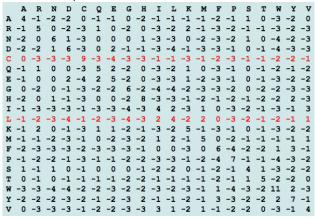
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- Alignments based on conventional scoring matrices (BLOSUM62) scores all positions in a sequence in an equal manner
- Some positions are highly conserved, some are highly variable (more than what is described in the BLOSUM matrix)
- Sequence profile are ideal suited to describe such position specific variations

Sequence alignment

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 Conventional sequence alignment uses a (Blosum) scoring matrix to identify amino acids matches in the two protein sequences

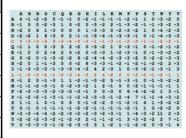


Alignment scoring matrices

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• Blosum62 score matrix. Fg=1. Ng=0?

	L	Α	G	D	5	D
F						
I						
G						
D						
5						
L						

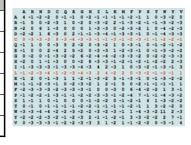


Alignment scoring matrices

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• Blosum62 score matrix. Fg=1. Ng=0?

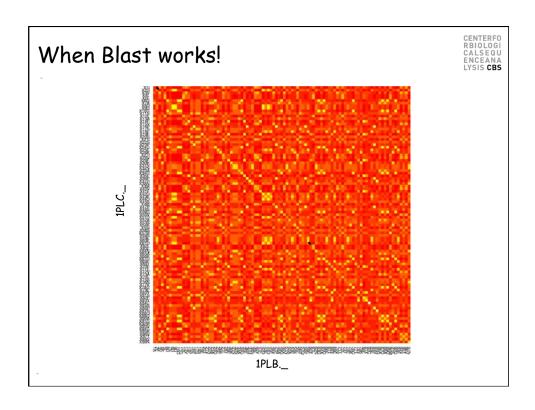
)
3
3
1
)
)
1
1



• Score =2-1+6+6+4=17

LAGDS

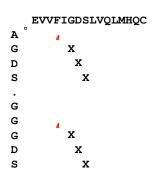
I-GDS



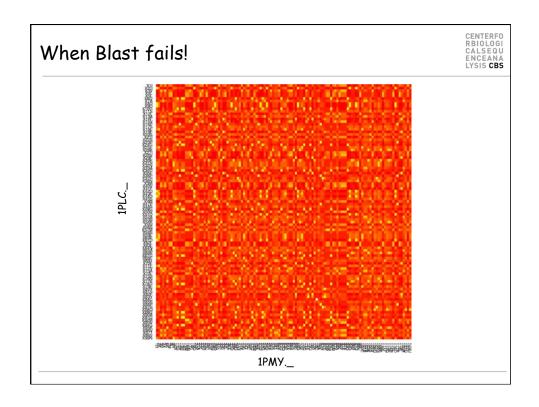
What goes wrong when Blast fails?

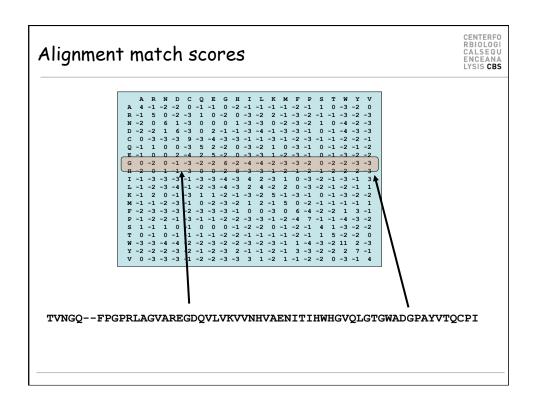
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- Conventional sequence alignment uses a (Blosum) scoring matrix to identify amino acids matches in the two protein sequences
- This scoring matrix is identical at all positions in the protein sequence!



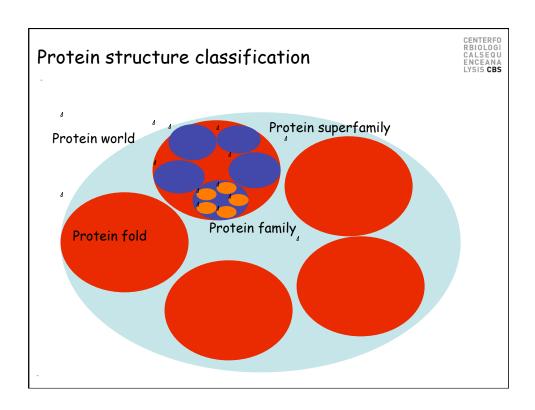
25





Sequence profiles

- In reality not all positions in a protein are equally likely to mutate
 - Some amino acids (active cites) are highly conserved, and the score for mismatch must be very high
 - Other amino acids can mutate almost for free, and the score for mismatch should be lower than the BLOSUM score
- Sequence profiles can capture these differences



Sequence profiles

 ${\tt TVNGQ--FPGPRLAGVARE} {\color{red}{\bf GD}} {\color{blue}{\bf QVLVKVVNHVAENITIHWHGVQLGTGWADP}} {\color{blue}{\bf PAYVTQCPI}}$ ${\tt TKAVVLTFNTSVEICLVMQ} {\color{red}{\bf G}} {\color{red}{\bf TSIV----AAESHPLHLHGFNFPSNFNLVD}} {\color{red}{\bf G}} {\color{red}{\bf MERNTAGVP}}$

Sequence profiles

Conserved

Non-conserved ADDGSLAFVPSEF--SISPGEKIVFKNNAGFPHNIVFDEDSIPSGVDASKISMSEEDLLN TVNGAI--PGPLIAERLKEGONVRVTNTLDEDTSIHWHGLLVPFGMDGVPGVSFPG---I -TSMAPAFGVQEFYRTVKQGDEVTVTIT----NIDQIED-VSHGFVVVNHGVSME---I IE--KMKYLTPEVFYTIKAGETVYWVNGEVMPHNVAFKKGIV--GEDAFRGEMMTKD----TSVAPSFSQPSF-LTVKEGDEVTVIVTNLDE----IDDLTHGFTMGHHGVAME---V ASAETMVFEPDFLVLEIGPGDRVRFVPTHK-SHNAATIDGMVPEGVEGFKSRINDE----TVNGQ--FPGPRLAGVAREGDQVLVKVVNHVAENITIHWHGVQLGTGWADPPAYVTQCPI TKAVVLTFNTSVEICLVMQGTSIV----AAESHPLHLHGFNFPSNFNLVDGMERNTAGVP

> Matching any thing but $G \Rightarrow large$ negative score

Any thing can match

$$p_a = \frac{\alpha \cdot f_a + \beta \cdot g_a}{\alpha + \beta} \qquad g_b = \sum_a f_a \cdot q_{b|a}$$

$$g_b = \sum_a f_a \cdot q_{b|a}$$

How to make sequence profiles

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- 1. Align (BLAST) sequence against large sequence database (Swiss-Prot)
- 2. Select significant alignments and make sequence profile
- 3. Use profile to align against sequence database to find new significant hits
- 4. Repeat 2 and 3 (normally 3 times!)

Sequence logos.

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Visualization of sequence profiles

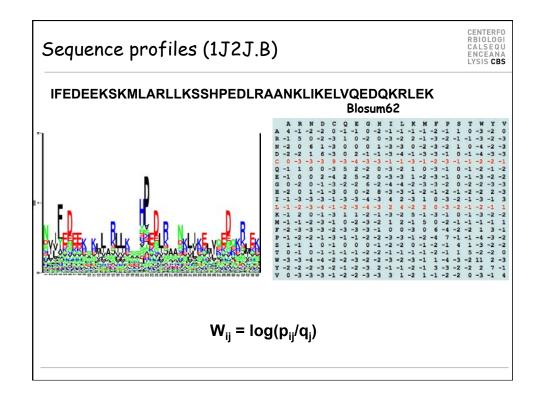
$$I = \sum_{a} p_a \log(\frac{p_a}{q_a})$$

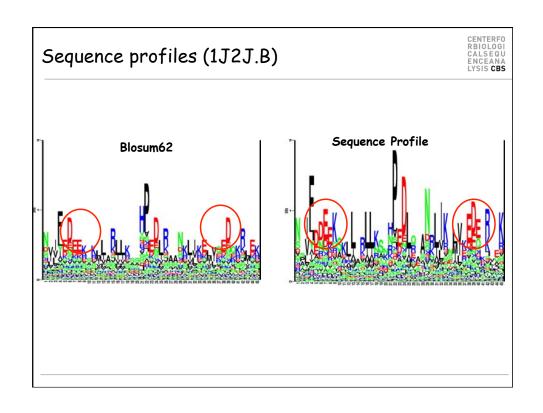
ALAKAAAAR
ALAKAAAAV
GMNERPILT
GILGFVFTM
TLNAWVKVV
KLNEPVLLL
AVVPFIVSV

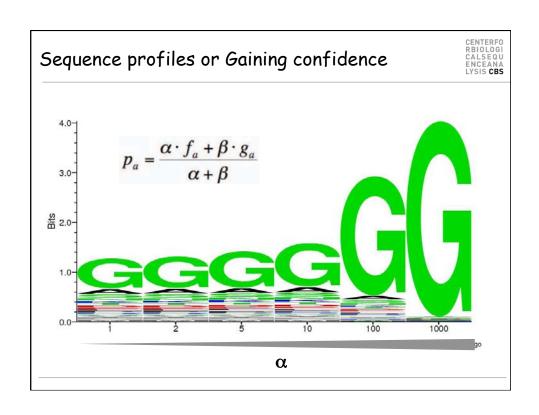
LAKAAAAM

$$\begin{array}{lll} P_A = 6/10 = 0.6 & q_A = 0.07 \\ P_G = 2/10 = 0.2 & q_G = 0.07 \\ P_T = P_K = 1/10 = 0.1 & q_T = 0.05 \\ P_C = P_D = ...P_V = 0.0 & q_K = 0.06 \end{array}$$

Sequence logos $I = \sum_a p_a \log(\frac{p_a}{q_a})$ High information positions $I = \sum_a p_a \log(\frac{p_a}{q_a})$ Height of a column equal to I Relative height of a letter is p (letters are upside down if q>p)







Example.

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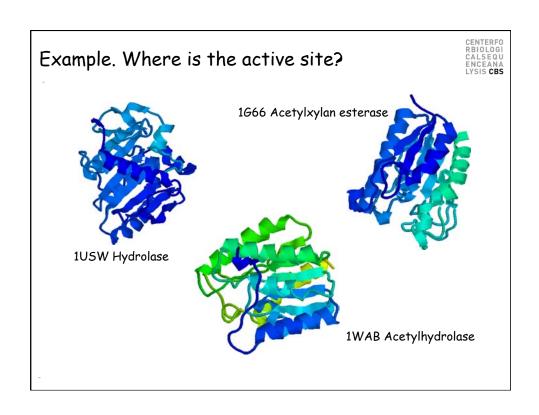
>1K7C.A

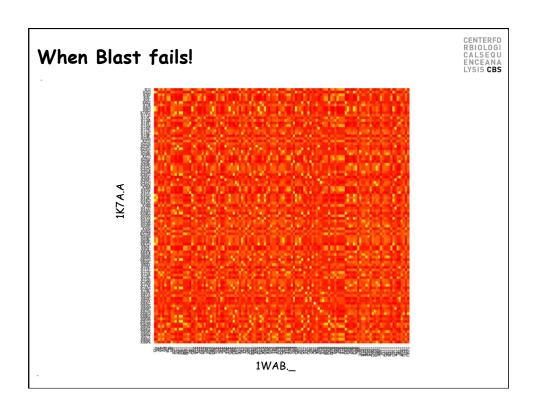
TTVYLAGDSTMAKNGGGSGTNGWGEYLASYLSATVVNDAVAGRSARSYTREGRFENIADV VTAGDYVIVEFGHNDGGSLSTDNGRTDCSGTGAEVCYSVYDGVNETILTFPAYLENAAKL FTAKGAKVILSSQTPNNPWETGTFVNSPTRFVEYAELAAEVAGVEYVDHWSYVDSIYETL GNATVNSYFPIDHTHTSPAGAEVVAEAFLKAVVCTGTSLKSVLTTTSFEGTCL

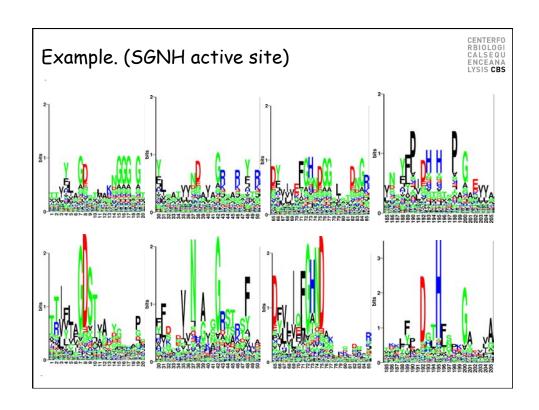
- What is the function
- Where is the active site?

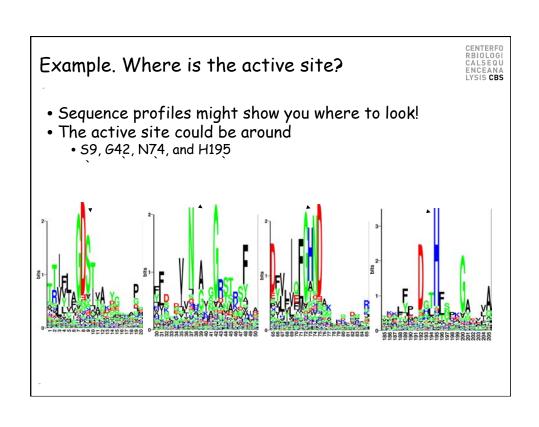
What would you do?

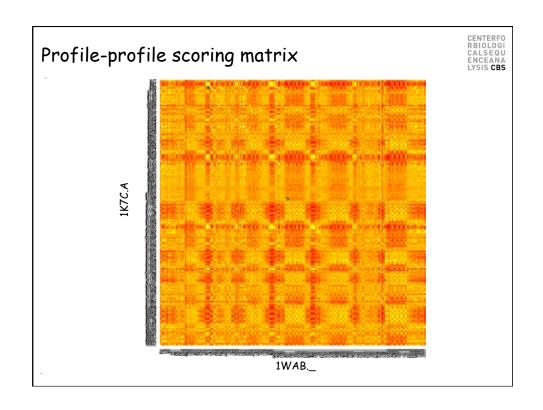
- Function
 - Run Blast against PDB
 - No significant hits
 - Run Blast against NR (Sequence database)
 - Function is Acetylesterase?
- Where is the active site?

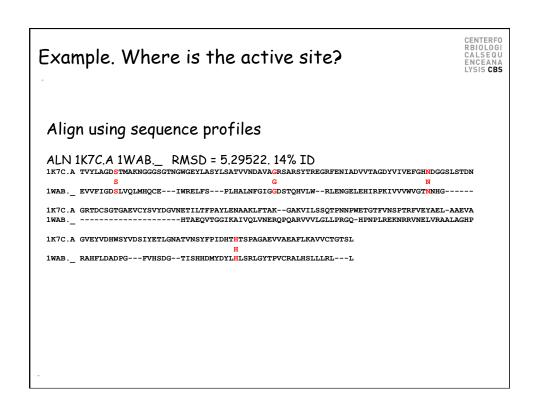






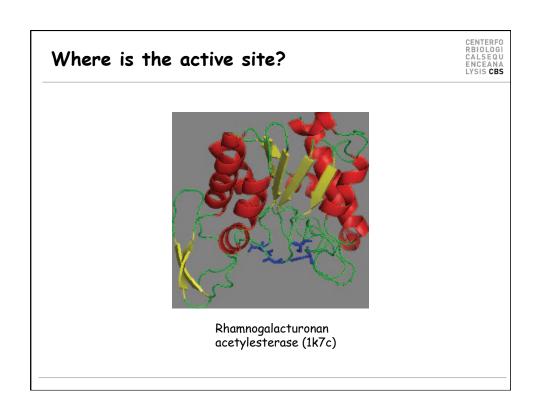


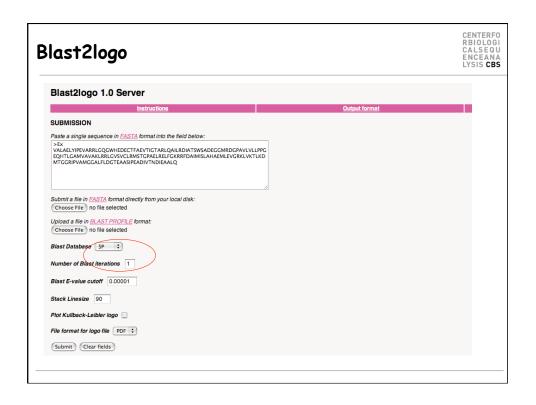


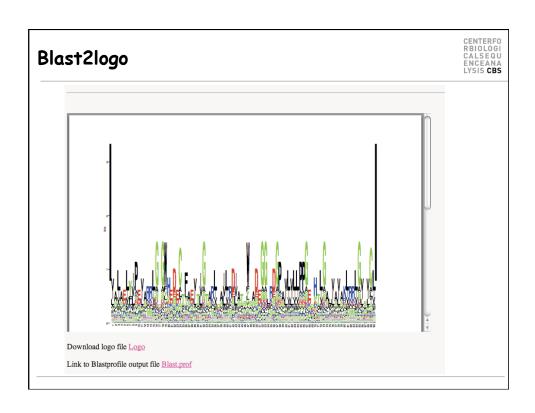


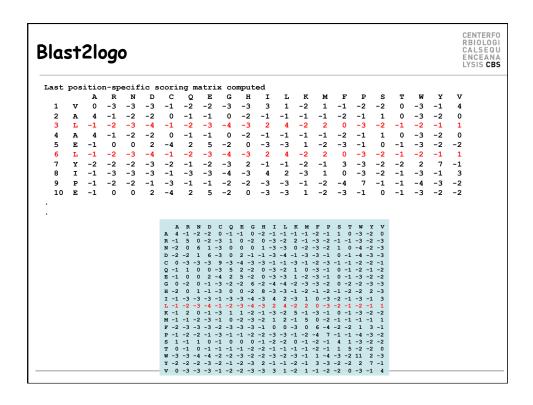
CENTERFO RBIOLOGI CALSEQU ENCEANA LYSIS CBS

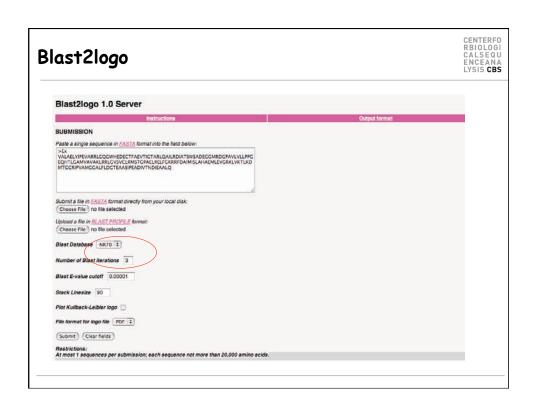
Handout exercise Using Psi-Blast Profiles

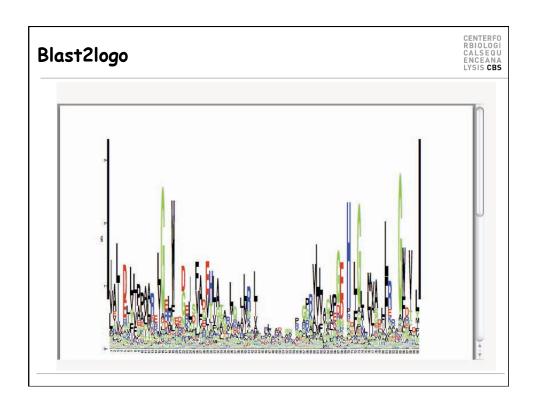


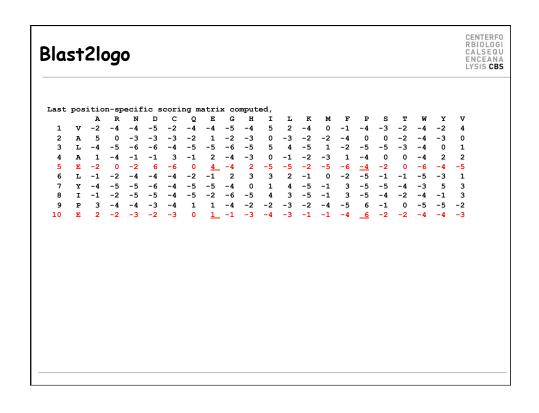












Sequence profiles take home message

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- Blast will often fail to recognize sequence relationships for low homology sequence pairs
- Sequence profiles contain information on conserved/ variable residues in a protein sequence
- Sequence profiles are calculated from (multiple) sequence alignments
- Iterative Blast enables homology recognition also for low sequence similarity
- Sequence profiles give information on residues essential for protein function and protein structure

Summary

- Sequence logo is a power tool to visualize (binding) motifs
 - Information content identifies essential residues for function and/or structural stability
- Weight matrices and sequence profiles can be derived from very limited number of data using the techniques of
 - Sequence weighting
 - Pseudo counts
- Weight matrices and sequences profiles can accurately describe binding motifs, sequence conservation, active sites...

The Beauty of Sequence profiles

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1)
$$\alpha = N-1 = 0$$

2)
$$p_a = q_a$$

2)
$$p_a = g_a$$

3) $f_G = 1$, $f_{|G} = 0$

4)
$$p_R = f_G * q(R|G) = 0.02$$

5)
$$q_R = 0.052$$

6)
$$Log$$
-odd = $2*log(p_a/q_a)/log(2) = -2.7$

7) Blosum62(
$$G_{1}$$
R) = -2

 $\texttt{TKAVVLTFNTSVEICLVMQ} \underline{\textbf{G}} \\ \texttt{TSIV----AAESHPLHLHGFNFPSNFNLVDPMERNTAGVP} \\$

$$p_a = \frac{\alpha \cdot f_a + \beta \cdot g_a}{\alpha + \beta} \qquad g_b = \sum_a f_a \cdot q_{b|a}$$

$$g_b = \sum_a f_a \cdot q_{b|a}$$

The Blosum matrix

ARNDCQEGHILKMFPS 4 -1 -2 -2 0 -1 -1 0 -2 -1 -1 -1 -1 -2 -1 1 0 -3 -2 R-1 5 0-2-3 1 0-2 0-3-2 2-1-3-2-1-1-3-2-3 $\verb|N-2 0 6 1-3 0 0 0 1-3-3 0 -2-3-2 1 0-4-2-3 | \\$ -2 -2 1 6 -3 0 2 -1 -1 -3 -4 -1 -3 -3 -1 0 -1 -4 -3 -3 0 -3 -3 -3 9 -3 -4 -3 -3 -1 -1 -3 -1 -2 -3 -1 -1 -2 -2 -1 Q -1 1 0 0 -3 5 2 -2 0 -3 -2 1 0 -3 -1 0 -1 -2 -1 -2 E -1 0 0 2 -4 2 5 -2 0 -3 -3 1 -2 -3 -1 0 -1 -3 -2 -2 G 0 -2 0 -1 -3 -2 -2 6 -2 -4 -4 -2 -3 -3 -2 0 -2 -2 -3 -3 H -2 0 1 -1 -3 0 0 -2 8 -3 -3 -1 -2 -1 -2 -1 -2 -2 2 -3 I -1 -3 -3 -3 -1 -3 -3 -4 -3 4 2 -3 1 0 -3 -2 -1 -3 -1 3 L -1 -2 -3 -4 -1 -2 -3 -4 -3 2 4 -2 2 0 -3 -2 -1 -2 -1 F -2 -3 -3 -3 -2 -3 -3 -3 -1 0 0 -3 0 6 -4 -2 -2 1 3 -1 -1 -2 -2 -1 -3 -1 -1 -2 -2 -3 -3 -1 -2 -4 7 -1 -1 -4 -3 -2 1 -1 1 0 -1 0 0 0 -1 -2 -2 0 -1 -2 -1 4 1 -3 -2 -2 W -3 -3 -4 -4 -2 -2 -3 -2 -2 -3 -2 -3 -1 1 -4 -3 -2 11 2 -3 Y -2 -2 -2 -3 -2 -1 -2 -3 2 -1 -1 -2 -1 3 -3 -2 -2 2 7 -1 V 0 -3 -3 -3 -1 -2 -2 -3 -3 3 1 -2 1 -1 -2 -2 0 -3 -1 4