Multiple Sequence Alignment

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Why to use MSA?

- To identify common conserved sequential motives and assess probability of their functional importance
- To obtain information about evolutionary relationships and history
- To construct phylogenetic trees

Pairwise Alignment

- Aligns two sequences
- We use dynamic programming
- Can be computed in O(nm)
- Parameters: gap penalties, substitution matrix
- We fill the matrix, always using maximum of three previously computed values:

$$f_{i, j} = \max\{f_{i-1, j-1} + s(a_i, b_j), f_{i-1, j} + gap, f_{i, j-1} + gap\}$$

Pairwise Alignment



 For each cell compute maximum of three neighbouring cells **3-D** Alignment



• For each cell compute maximum of seven neighbouring "cubes"

3-D Alignment



Where s is a 3-dimensional substitution matrix

k-D Alignment

- Assume we want to align k sequences, each n symbols long. We need to fill a k-dimensional array, thus running time is O(n^k).
- Because of exponential running time, we don't usually use k-dimensional multiple alignment
- Although this can be improved by Carrilo-Lipman Heuristic which sets a bound of the score of alignment so that not all regions of the dynamic programming lattice have to be explored

Back to pairwise Alignment

- Can we align more than two sequences using only pairwise alignment?
- Idea: assume two aligned sequences, we will call it a profile. We can easily extend the pairwise alignment to work with profiles

AT-AGTTC + TTGAGTC = AT-AGTTCTTGAGT-C

Aligning Profiles

	A A	T T	G_	A A	G _G	A T	T T	с С
Т								
G								
Α								
G								
Т								
Α								
С								

 The algorithm works similarly, but computing the substitution value is a little different

$$S(a_i, b_j) = \sum_x (P_i(x) \sum_y (P_j(y) \cdot s(x, y)))$$

Progressive Multiple Alignment

- In what order should we add sequences to the profile?
- Generally, a tree model is preferred as it is biologically most relevant. First align most similar sequences and then add them to the rest of the sequences.
- We will need a similarity matrix

Similarity matrix

	s1	s2	s3	s4	s5	s6
s1	-	-	-	-	-	-
s2	0.17	-	-	-	-	-
s3	0.59	0.60	-	-	-	-
s4	0.59	0.59	0.13	-	-	-
s5	0.77	0.77	0.75	0.75	-	-
s6	0.81	0.82	0.73	0.74	0.80	-

- At each step we combine two most similar clusters.
- Similarity of two clusters A and B is defined as an average of similarities of pairs of sequences in A and B

$$S(A, B) = \frac{1}{|A| \cdot |B|} \sum_{x \in A} \sum_{y \in B} s(x, y)$$

 This method is called Unweighted Pair Group Method with Arithmetic mean (UPGMA)

Dendrograms

- Are created by methods like UPGMA or Neighbour-joining.
- Concern an evolutionary distance of sequences
- Also called Guide Trees

Dendrograms - example



Example

gi 4557040	MALFAVFQTTFFLTLLSL R TYQS <mark>EVLAER</mark> LPLTPVSLKVSTLSTRQSLHLQWTVHNLPYHQ <mark>ELK</mark> MVFQ
gi 119576380	MALFAVFQTTFFLTLLSL R TYQSEVLAERLPLTPCVSL R VSTNST R QSLHLQUTVHNLPYHQELKMVFQ
gi 109659086	M L T L Q T U L V Q A L F I F L T T E S T G E L L D P C G Y I S P E S P V V Q L H S N F T A V C V L K E K C M D Y F H V N A N Y I V U K
gi 3153816	MALFSVVLHPAFLLAVLSL RASRSEVLEEPLPLTPEIHKVSFQLKLQEVNLEUTVPALTHEELNMIFQI
gi 261858134	MALFAVFQTTFFLTLL <mark>SLR</mark> TQSEVLAERL <mark>PLTPVSLRVSTL</mark> ST <mark>R</mark> QSLHLQWTVHNLPYHQELKNVFQ
gi 162287202	MAFS VVLHPAFLLAVLSLRASR SEVFEEPLPLTPEIHKVSFQLKLQEVNLEWTVPALTHEELNNIFQI
gi 119575331	M L T L Q T U L V Q A L F I F L T T E S T G E L L D P C G Y I S C E S P V V Q L H S N F T A V C V L K E K C M D Y F H V N A N Y I V U K
gi 52851389	MAFSVVLHQVTFLLAVLSL R TSQS K VL <mark>GEPLQLTPEIHTVSLQSALQEANLEWTV</mark> PTFSHQ <mark>ELNIVF</mark> Q
gi 223460974	MAFS VVLHPAFLLAVLSLRASR SEVLEEPLPLTPEIHKVSFQLKLQEVNLEWTVPALTHEELNNIFQI
gi 148232174	M L T L Q T U V V Q A L F I F L T T K C K G E L L D P C G H I S P E S P V I Q L G S N F T A V C V L K E K C M D H Y H V N A S Y I F U K





Progressive Multiple Alignment



gi 148232174	MLTLQTWVWQALFIFLTTKCKGE-L-LDPCGHISPESPVIQ-LGSNFTAWCVLKEKCMDHYHWNASYIFW	Κ
[─[_gi 109659086	MLTLQTWLVQALFIFLTTESTGE-L-LDPCGYISPESPVVQ-LHSNFTAVCVLKEKCMDYFHVNANYIVW	K
∟gi 119575331	MLTLQTULVQALFIFLTTESTGE-L-LDPCGYISCESPVVQ-LHSNFTAVCVLKEKCMDYFHWNANYIVU	K
gi 119576380	MALFAV FOTTFFLTLLSL <mark>R</mark> TYOSEVLAERLPLTPCVSLRVSTNS-TROSLHLOWT-VHNLPYHOELKMVFO	-
┌┤_gi 4557040	MALFAV FOTTFFLTLLSL <mark>R</mark> TYOSEVLAERLPLTP - VSLKVSTLS-TROSLHLOWT-VHNLP YHOELKMVFO	-
Ц └gi 261858134	MALFAV FOTTFFLTLLSL <mark>R</mark> T - OS <mark>EVLAERLPLTP - VSLRVSTLS - TROSLHLOWT - VHNLP</mark> YHQ ELKMVF Q	-
gi 52851389	MA-FSVVLHQWTFLLAVLSL <mark>R</mark> TS <u>Q</u> SKVLGEPLQLTPEIH-TVSLQ <u>S</u> -ALQEANLEMT-VPTFSHQELNIVFQ	-
µgi 162287202	MA – FSVVLHP – AFLLAVLSL <mark>R</mark> AS R SEVFEEPLPLTPEIH – KVSFQL – KLQEVNLEMT – VPALT – – – – – – HEELNMIFQ	Ι
ე_gi 3153816	MALFSVVLHP-AFLLAVLSL <mark>R</mark> AS R SEVLEEPLPLTPEIH-KVSFQL-KLQEVNLEMT-VPALTHEELNMIFQ	Ι
└gi 223460974	MA – FSWVLHP – AFLLAVLSLRASRSEVLEEPLPLTPETH – KWSFQL – RLQEWNLEMT – WPALT – – – – – – HEELNMIFQ	Ι
gi 52851389 gi 162287202 gi 3153816 gi 223460974	MA – FSVVLHQMTFLLAVLSLRTSQSKVLGEPLQLTPEIH – TVSLQS – ALQEANLEWT – VPTFS – – – – – – HQELNIVFQ MA – FSVVLHP – AFLLAVLSLRASRSEVFEEPLPLTPEIH – KVSFQL – KLQEVNLEWT – VPALT – – – – – – HEELNMIFQ MALFSVVLHP – AFLLAVLSLRASRSEVLEEPLPLTPEIH – KVSFQL – KLQEVNLEWT – VPALT – – – – – – HEELNMIFQ MA – FSVVLHP – AFLLAVLSLRASRSEVLEEPLPLTPEIH – KVSFQL – KLQEVNLEWT – VPALT – – – – – – HEELNMIFQ	– I I I

Progressive Multiple Alignment

- Once we have inserted a gap into sequence, it stays there
- Therefore we have to build strong initial alignments
- Clustal, T-Coffee

ClustalW

- Distance Matrix (Pairwise Alignments)
- Guide Tree
- Progressive Alignment
- Gap Open Penalty, Gap Extension Penalty
 - Similarity of sequences
 - Lengths of sequences
 - "GOP->(GOP+log(MIN(N,M))) * (average residue mismatch score) * (percent identity scaling factor)"
 - "GEP -> GEP*(1.0+|log(N/M)|)"
- 80-100%: PAM20, 60-80%: PAM60, 40-60%: PAM120, 0-40%: PAM350.
- 80-100%: BLOSUM80, 60-80%: BLOSUM62, 30-60%: BLOSUM45, 0-30%: BLOSUM30

ClustalW



Enter or paste a set of seq	uences in any supported format:	Help
Upload a file:	Browse_	Run Reset

Iterative Multiple Alignment

- When constructing alignment, it realigns sequences already aligned
- Variety of methods exists
- For example: after the alignment is done, remove a sequence and add it to the alignment again
- MUSCLE (multiple sequence comparison by log-expectation)

Other methods

- Many other methods have been used to align more sequences
- Hidden Markov Models, Motif finding, Genetic algorithms

- How to find out which alignment is better?
- How do we mathematically define "better"?
- Sum of Pairs Score:

$$SP \begin{pmatrix} ATC-TAC \\ ATC-TAG \\ A-CCTTG \\ A-CGTTG \end{pmatrix} = SP(AAAA) + SP(TT--) + SP(CCCC) + SP(--CG) + SP(CCCC) + SP(--CG) + SP(TTTAA) + SP(TTTAA) + SP(CGGG)$$

$$SP(--CG) = s(-,-) + s(-,C) + s(-,G) + s(-,C) + s(-,C) + s(-,G) + s(C,G)$$

• Entropy:

$$Entropy = \sum_{all \ columns} \sum_{x \in Alphabet} p_x \cdot \log(p_x)$$

Alignments with lower entropy are better

 Comparing our own method with Clustal using Entropy objective function

Protein	We	Clustal
ccl2	36.89	39.17
cd147	116.43	117.39
cd154	177.89	179.49
Collagen alpha	659.50	656.31
prolactin	248.77	241.32
resistin	58.65	59.40
selectinL	114.95	114.95



 Comparing our own method with Clustal using Sum-of-Pairs objective function (Blosum62)

Protein	We	Clustal
ccl2	35686	35782
cd147	5279	5255
cd154	34064	36011
Collagen alpha	78360	78534
prolactin	32432	52804
resistin	5064	5057
selectinL	9481	9481

