Multiple Sequence Alignment

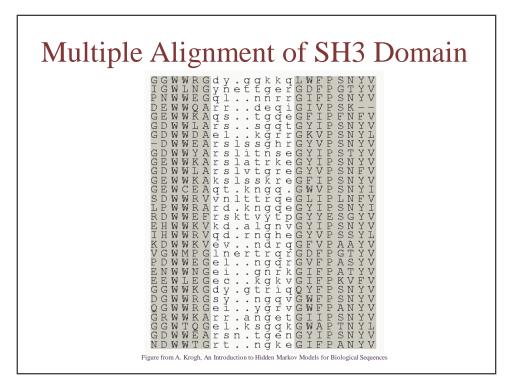
BMI/CS 576 www.biostat.wisc.edu/bmi576.html Colin Dewey cdewey@biostat.wisc.edu

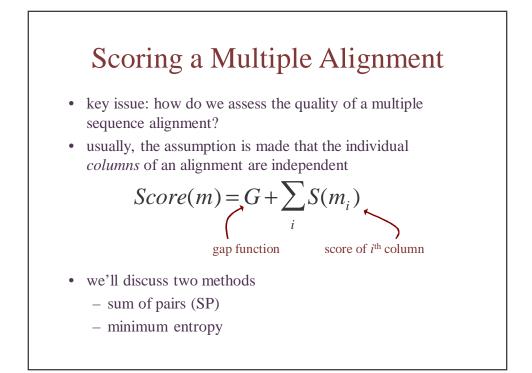
Multiple Sequence Alignment: Task Definition

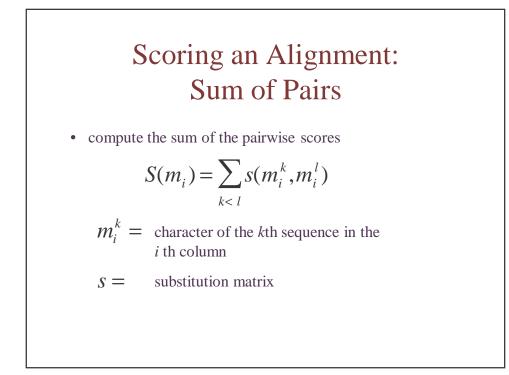
- Given
 - a set of more than 2 sequences
 - a method for scoring an alignment
- Do:
 - determine the correspondences between the sequences such that the alignment score is maximized

Motivation for MSA

- establish input data for phylogenetic analyses
- determine evolutionary history of a set of sequences
 At what point in history did certain mutations occur?
- discovering a common motif in a set of sequences (e.g. DNA sequences that bind the same protein)
- characterizing a set of sequences (e.g. a protein family)
- building *profiles* for sequence-database searching
 - PSI-BLAST generalizes a query sequence into a profile to search for remote relatives







Scoring an Alignment: Minimum Entropy

- basic idea: try to minimize the entropy of each column
- another way of thinking about it: columns that can be communicated using few bits are good
- information theory tells us that an optimal code uses $-\log_2 p$ bits to encode a message of probability p

Scoring an Alignment: Minimum Entropy

- the messages in this case are the characters in a given column
- the entropy of a column is given by:

$$S(m_i) = -\sum_a c_{ia} \log_2 p_{ia}$$

 $m_i =$ the *i* th column of an alignment *m*

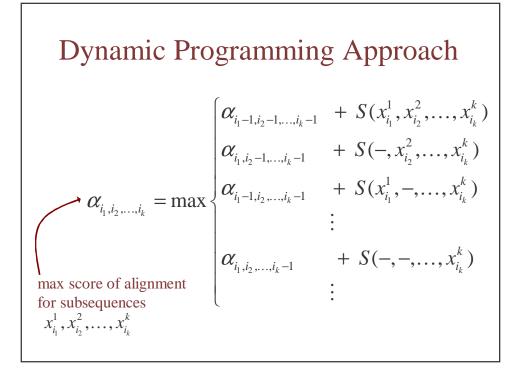
 $c_{ia} = \text{count of character } a \text{ in column } i$

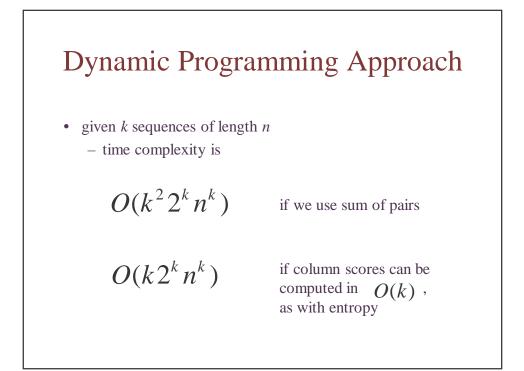
 $p_{ia} =$ probability of character *a* in column *i*

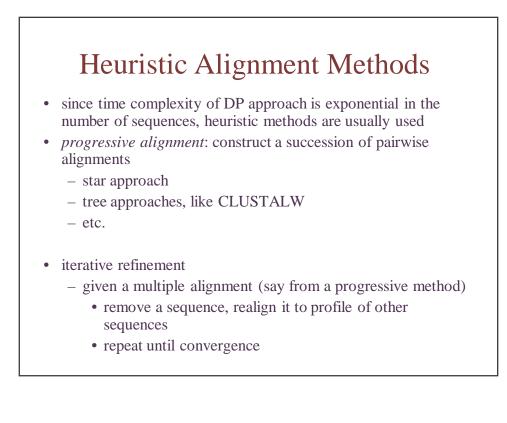
Dynamic Programming Approach

- can find optimal alignments using dynamic programming
- · generalization of methods for pairwise alignment
 - consider *k*-dimension matrix for *k* sequences (instead of 2-dimensional matrix)
 - each matrix element represents alignment score for k subsequences (instead of 2 subsequences)
- given k sequences of length n
 - space complexity is

 $O(n^k)$







Star Alignment Approach

• given: *k* sequences to be aligned

 $X_1, ..., X_k$

- pick one sequence X_c as the "center"
- for each $X_i \neq X_c$ determine an optimal alignment between X_i and X_c
- merge pairwise alignments
- return: multiple alignment resulting from aggregate

Star Alignments: Approaches to Picking the Center

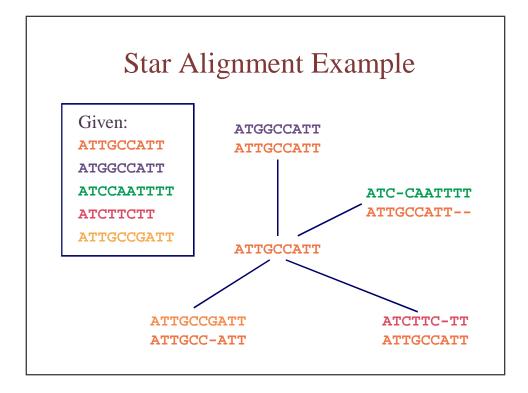
Two possible approaches:

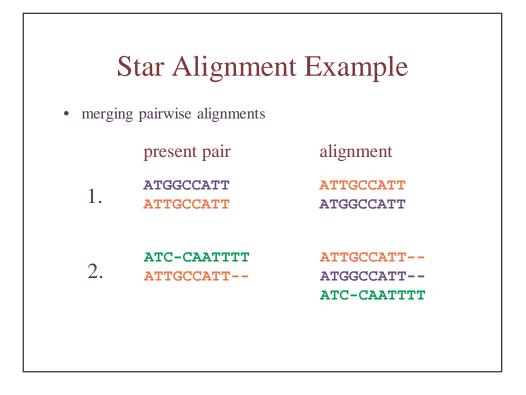
- 1. try each sequence as the center, return the best multiple alignment
- 2. compute all pairwise alignments and select the string x_c that maximizes:

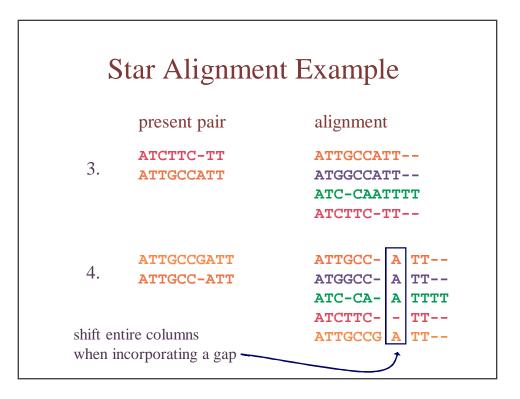
$$\sum_{i\neq c} \operatorname{sim}(x_i, x_c)$$

Star Alignments: Aggregating Pairwise Alignments

- "once a gap, always a gap"
- shift entire columns when incorporating gaps







Tree Alignments

- basic idea: organize multiple sequence alignment using a *guide tree*
 - leaves represent sequences
 - internal nodes represent alignments
- determine alignments from bottom of tree upward
 - return multiple alignment represented at the root of the tree
- one common variant: the CLUSTALW algorithm [Thompson et al. 1994]

Doing the Progressive Alignment in CLUSTALW

- depending on the internal node in the tree, we may have to align a
 - a sequence with a sequence
 - a sequence with a *profile* (partial alignment)
 - a *profile* with a *profile*
- in all cases we can use dynamic programming
 - for the profile cases, use SP scoring

