Pairwise Sequence Alignment

Manu Madhavan

Lecture 5

Manu Madhavan 15CSE335 Lecture 5 1/23

Outline

- What is Sequence Alignment
- Importance of Sequence Alignment
- Global and Local Alignment
- Algorithms for Global Alignment

Manu Madhavan 15CSE335 Lecture 5 2/21

Biological Problem

Sequence alignment is a way of arranging protein (or DNA) sequences to identify regions of similarity that may be a consequence of evolutionary relationships between the sequences.

- Genome sequencing allows comparison of organisms at DNA and protein levels
- Comparisons can be used to
 - Find evolutionary relationships between organisms
 - Identify functionally conserved sequences
 - Identify corresponding genes in human and model organisms: develop models for human diseases

Sequence Homology

- Homology: genes that derive from a common ancestor-gene are called homologs
- Orthologous genes are homologous genes in different organisms
- Paralogous genes are homologous genes in one organism that derive from gene duplication
- Gene duplication: one gene is duplicated in multiple copies that therefore free to evolve and assume new functions

Manu Madhavan 15CSE335 Lecture 5

Sequence similarity

- Intuitively, similarity of two sequences refers to the degree of match between corresponding positions in sequence
- Sequence similarity is not sequence homology
- Homology is more difficult to detect over greater evolutionary distances

Causes of Gene (dis) similarity

- mutation: a nucleotide at a certain location is replaced by another nucleotide. ATA → AGA
- insertion: at a certain location one new nucleotide is inserted in between two existing nucleotides (e.g.: AA → AGA)
- **deletion**: at a certain location one existing nucleotide is deleted (e.g.: ACTG \rightarrow AC-G)
- indel: an insertion or a deletion

Manu Madhavan 15CSE335 Lecture 5 6/21

- Find the similarity between two (or more) DNA-sequences by finding a good alignment between them
- Alignment specifies which positions in two sequences match



- Sequence alignment is an arrangement of two or more sequences, highlighting their similarity.
- The sequences are padded with gaps (dashes) so that wherever possible, columns contain identical characters from the sequences involved

```
tcctctgcctctgccatcat---caaccccaaagt
```

- Pairwise Sequence Alignment: methods are concerned with finding the best-matching piece-wise local or global alignments of protein (amino acid) or DNA (nucleic acid) sequences.
- Global Alignment: an alignment in which all the characters in both sequences participate in the alignment.
- Local Alignment: a matching two sequence from regions which have more similar with each other
- Multiple Alignment

Manu Madhavan 15CSE335 Lecture 5 9 / 21

Algorithms

- Needleman-Wunsch
 Pairwise global alignment only.
- Smith-Waterman
 Pairwise, local (or global) alignment.
- BLAST
 Pairwise heuristic local alignment

Manu Madhavan 15CSE335 Lecture 5 10 / 21

The Needleman-Wunsch algorithm

- The Needleman-Wunsch algorithm (1970, J Mol Biol. 48(3):443-53) performs a global alignment on two sequences (s and t) and is applied to align protein or nucleotide sequences.
- The Needleman-Wunsch algorithm is an example of dynamic programming, and is guaranteed to find the alignment with the maximum score.

Manu Madhavan 15CSE335 Lecture 5 11/21

The Needleman-Wunsch algorithm

Alignment Scoring Function

The cost of aligning two symbols x_i and y_j is the scoring function $\sigma(x_i, y_j)$

$$\sigma(x_i, y_j) = \begin{cases} -1 & \text{if } x_i \neq y_j \text{ or } (x_i, _) \text{ or } (_, y_j) \\ 1 & \text{if } x_i = y_j \end{cases}$$

More better scores can be calculated by PAM (Point Accepted Mutation) matrix (Margaret Dayhoff), BLOSUM (BLOck SUbstitution Matrix) (Henikoff and Henikoff)

Alignment cost

$$M = \sum_{i=1}^{c} \sigma(x_i, y_j)$$

Dynamic Programming

- A matrix D(i,j)indexed by residues of each sequence is built recursively,such that
- A gap from left or above cell
- A match/mismatch from diagonal element

$$D(i,j) = max egin{cases} D(i-1,j-1) + s(x_i,y_j) \ D(i-1,j) + g \ D(i,j-1) + g \end{cases}$$

where $s(x_i, y_j)$ is the substitution cost g is the gap penalty

Manu Madhavan 15CSE335 Lecture 5 13 / 21

Dynamic Programming-steps

- Initialization of the score matrix
- Calculation of scores and filling the traceback matrix
- Deducing the alignment from the traceback matrix

Manu Madhavan 15CSE335 Lecture 5 14 / 21

Let's work on this simple example

• Input: GACTT (sequence #1), ATT (sequence #2)

Manu Madhavan 15CSE335 Lecture 5 15 / 21

Steps

ullet Input: CGTGAATTCAT (sequence #1), GACTTAC (sequence #2)

	_	C	G	T	G	A	A	T	T	C	A	T
-				6						- 6		0
G												
A				100		7						
C												
T												
T				100						- 3		6
A												
C												

Manu Madhavan 15CSE335 Lecture 5 1

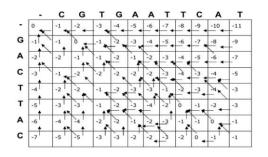
Steps: Initialization

ullet Input: CGTGAATTCAT (sequence #1) , GACTTAC (sequence #2)

	_	C	G	T	G	A	A	T	T	C	A	Т
-	0	-1	-2	-3	-4	-5	-6	-7	-8	-9	-10	-11
G	-1			\vdash								
A	-2											
С	-3		T							Т		
т	-4											
т	-5		\vdash									
A	-6											
С	-7											

Steps: Filling the matrix

• Input: CGTGAATTCAT (sequence #1), GACTTAC (sequence #2)

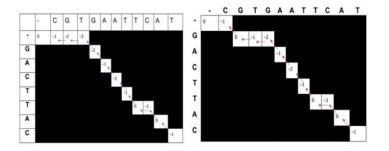


$$D(i,j) = max egin{cases} D(i-1,j-1) + s(x_i,y_j) \ D(i-1,j) + g \ D(i,j-1) + g \end{cases}$$

where $s(x_i, y_j)$ is the substitution cost g is the gap penalty

Steps: Traceback

• Input: CGTGAATTCAT (sequence #1), GACTTAC (sequence #2)



Manu Madhavan 15CSE335 Lecture 5 19 / 21

Exercise

• Input: ACAGTAG (sequence #1), ACTCG (sequence #2)



Manu Madhavan 15CSE335 Lecture 5 20 / 21

Next

Reading

Chapter-2, Krane and Raymer, Fundamentals of Bioinformatics

- Global Sequence Alignment
- BLAST
- MSA

Manu Madhavan 15CSE335 Lecture 5 21 / 21