References

CSI5126. Algorithms in bioinformatics Pairwise Sequence Alignment

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Version September 27, 2018

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In this lecture, we learn that molecular sequences suffer mutations. We distinguish between two kinds of similar sequences: orthologues and paralogues. We derive an algorithm to compare molecular sequences taking into account their mode of evolution. **General objective**

Describe in your own words the pairwaise sequence alignment problem and explains its asumptions.

Reading

 Bernhard Haubold and Thomas Wiehe (2006). Introduction to computational biology: an evolutionary approach. Birkhäuser Basel. Pages 11-15, 30-33.

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Comparative sequence analysis

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Why?

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Why?

"Determining **function** for a sequence is a matter of tremendous complexity, requiring biological experiments of the highest order of creativity. Nevertheless, with only DNA sequence it is possible to execute a computer-based algorithm comparing the sequence to a database of previously characterized genes. In about 50% of the cases, such a mechanical comparison will indicate a sufficient degree of **similarity** to suggest a putative enzymatic or structural function that might be possessed by the unknown gene."

Caskey et al. (1995) Genome Digest 2:6-9.

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A molecular sequence alignment aims

- to identify **similar regions** between two sequences
- to determine if two sequences have a **common origin**



Molecular sequences are the result of evolutionary processes.

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 Alignment
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 Comparative sequence analysis

- Molecular sequences are the result of evolutionary processes.
- Speciation (the formation of new and distinct species in the course of evolution) is the main process for creating new, yet related, sequences.

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- Information (function, structure, etc.) that is known about a sequence can generally be transferred to "similar" sequences.

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- Evolution transforms the sequences: point mutations (insertions, deletions, substitutions), duplications, inversions, transpositions, etc. Consequently, making it more difficult (interesting) to find the common origins.
- Information (function, structure, etc.) that is known about a sequence can generally be transferred to "similar" sequences.
- Comparative sequence analysis is therefore an **essential** and **powerful tool**.



All forms of life are believed to have evolved from a common origin.

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- **Conclusion**: all the sequences are related one to another.



- All forms of life are believed to have evolved from a common origin.
- The genomic content of the proto-cell (proto-organism) certainly arose by a series of events, including duplication content, from smaller sequence fragments.
- **Conclusion**: all the sequences are related one to another.
- This is not a very productive statement. The evolutionary relationships that are considered interesting are those that can be explained by the techniques presented here, for which there are convincing statistical evidences.

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Image: A matched block of the second seco

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Which metric is adequate to compare molecular sequences?

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- How to compute the alignment efficiently?

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- How to align sequences when $|S_1| << |S_2|$?
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- How to score insertions and deletions?



- Which metric is adequate to compare molecular sequences?
- How to compute the alignment **efficiently**?
- How to align sequences when $|S_1| \ll |S_2|$?
- How to score **substitutions**?
- How to score insertions and deletions?
- Any two sequences can be aligned, how to evaluate (the **likelihood** of) an alignment?



A VLSAADKGNVKAAWGKVGGHAAEYGAEALERMFLSFPTTKTYFPHFDLSHGSAQVKG B SLSAAQKDNVKSSWAKASAAWGTAGPEFFMALFDAHDDVFAKFSGLFSGAAKGTVKN

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14 out of 57 (25 % of) amino acids are identical.



Insertion/deletions, two evolutionary events, must be taken into account

21 out of 60 (35 % of) positions are identical.



Indel stands for insertion or deletion.

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- Given exactly two sequences, I am claiming that insertions cannot be distinguished from deletions, hence the use of the word indel.

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- Indel stands for insertion or deletion.
- Given exactly two sequences, I am claiming that insertions cannot be distinguished from deletions, hence the use of the word indel. What do I mean?
- Consider the following pairwise alignment, was the U, present in S1, deleted, to produce S2? Or, was is a U inserted into S2 to produce S1?
- S1 = UGCUUA
- S2 = UGC-UA









38 out of 60 (63 % of) positions have the same or similar properties.

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 Requirements
 Are these two sequences similar? (...have similar regions?)

> Methanococcus vannielii SB, DNA-directed RNA polymerase ATGGATAGATTTGATGTTCCAAAGGAAATCGGAGATATTACATTTGGATTGCTCTCTCCA GAACAGATAAGGACAATGTCTGTTGCAAAAATCGTTACAGCAGATACTTATGATGACGAT ... (2,670) ... ACAAAAGTCATTTCAAAATATGAAAATTAA

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- A dot plot is a useful tool to compare two sequences.
- It consists of a two dimensional diagram, such that one sequence is written along one of its axes, and the other sequence along the other axis.
- In its simplest form, a dot is plotted at position *i* and *j* if the characters *i* and *j* of the two respective strings are identical.

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The resulting diagram often contain too much noise (is too busy).


A **window-based** approach is generally used to circumvent the problem, i.e. a **dot** is plotted only if *x* **characters** (amino acids or nucleotides) **out** *w* **characters** are **identical**, where *w* is the window size.

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Insertions/deletions show up as slightly shifted diagonal



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- Shows duplications



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- Shows duplications
- Identifies local similarity
- Shows inverted repeats (anti-diagonals)
- Not suitable for automated analyses

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When insertions and deletions are allowed, there are many possible alignments of the two input sequences.

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- S1 A T T C G
 S1 A T T C G S1 A T T C G

 S2 T T C C A
 S2 T T C C A
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 x
 x
 x x
 - How many alignments for two input sequences of length 5?

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 - How many alignments for two input sequences of length 5? 1,683

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 x
 x
 x x x
 - How many alignments for two input sequences of length 5? 1,683
 - Which one to choose?
 - The edit distance is the minimum number of edit operations that are needed to transform one string into the other.

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 S1 A T T C G
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 S2 - T T C C A
 S2 - T T C - C A

 x
 x
 x x x

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- Which one to choose?
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The edit distance is sometimes referred to as **Levenshtein distance**.



The edit operations that are useful to model evolutionary processes are **insertions** (I), **deletions** (D) and the **substitutions** (S).

S1	А	Т	Т	С	G	
S2		Т	Т	С	С	A

- The set of operations can be augmented with the match (M) operation, which simply rewrites a letter from the input onto the output.
- However, the match operation will not be counted when calculating the edit distance; in other words, it can be seens as having a weight of 0.

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What are the assumptions?

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- What are the assumptions?
- Independent. These operations are independent one from another. The likelihood of a substitution at position *i* is not affected by the identity of the residue found at position *j*. Is this realistic?



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- Is this realistic?



An **edit transcript** is a string over $\{I, D, S, M\}$ that summarizes the edit operations that are applied to the first string in order to produce the second one.

 $\begin{array}{rrrr} Transcript & : & D M M M S I \\ S1 & : & A T T C G - \\ S2 & : & - T T C C A \end{array}$



- A string alignment consists of two input strings, written one on the top of the other, such that space (or dash) symbols have been added to the first, or second, string when insertions, or deletions, are seen in the edit transcript.
- The edit distance problem consists in finding the alignment (or equivalently the edit transcript) that minimizes the edit distance.



Size of the search space



Logarithm of the number of alignments as a function the sequence length

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- Uses of the edit distance occur outside of the context of biological sequence comparisons, examples are: spelling correction methods or textual database retrieval.
- The Unix program diff is an example of a program that is based on the notion of edit distance. It is a program that compares the content of two files.
- When ran with the argument **e** the program program produces a series of commands for the editor **ed** to transform the first file into the other.

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- I like considering the problem from the point of view of the edit transcript.

The edit transcript of the optimal alignment will **end** with one of the four edit operations, **I**, **D**, **S** or **M**.

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$$egin{array}{ccc} & I \ S_1 & - \ S_2[1,m-1] & S_2(m) \end{array}$$

where $S_2(m)$ is the last symbol of S_2 . Make sure to understand the details of above illustration. S_2 has been decomposed into a prefix and the last symbol, $S_2[1, m-1]S_2(m) = S_2$, a dash symbol has been added to the end of S_1 .

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 Assuming this transcript leads to an optimal alignment, how many edit operations are needed to transform S₁ into S₂?

Similarly for D (deletion),

$$\begin{array}{ccc} & D \\ S_1[1, n-1] & S_1(n) \\ S_2 & - \end{array}$$

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and S (substitution),

$$egin{array}{ccc} S \ S_1[1,n-1] & S_1(n) \ S_2[1,m-1] & S_2(m) \end{array}$$

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and M (match),

$$\begin{array}{c} M \\ S_1[1, n-1] & S_1(n) \\ S_2[1, m-1] & S_2(m) \end{array}$$

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Obviously, only if $S_1(n) = S_2(m)!$

and M (match),

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Obviously, only if $S_1(n) = S_2(m)!$

For each case, how many edit operations are needed to transform S₁ into S₂?

$$S_1 = \frac{I}{S_2[1, m-1]} = S_2(m)$$

The number of edit operations required is?


$$S_{1} = \frac{I}{S_{2}[1, m-1]} = S_{2}(m)$$

$$D(S_1, S_2[1, m-1]) + 1$$

$$\begin{array}{cc} & D\\ S_1[1,n-1] & S_1(n)\\ S_2 & -\end{array}$$



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$$M \\ S_1[1, n-1] \quad S_1(n) \\ S_2[1, m-1] \quad S_2(m)$$

$$D(S_1[1, n-1], S_2[1, m-1]) + 0$$

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- Consider $S_1 = \underline{ATT}GC$, $S_2 = \underline{A}GC$, and D(3, 1), it does not mean that $S_1(3) = T$ is aligned against $S_2(1) = A$

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- Any other alignment than the one below would involve 3 or more edit operations (2 deletions and one substitution)

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S1 ATT | S2 A--

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 - S1 ATT

S2 A--

 Here, the edit transcript of the optimal alignment is ending with a deletion (D)

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Let's see if we can find some base conditions.

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Let's see if we can find some base conditions. D(0,0) = ?

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Let's see if we can find some base conditions. D(0,0) =?
Surely, D(0,0) = 0,

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- Let's see if we can find some base conditions. D(0,0) = ?
- Surely, D(0,0) = 0, no operations are needed to transform the first zero characters of S₁ into the first zero characters of S₂.

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- Let's see if we can find some base conditions. D(0,0) = ?
- Surely, D(0,0) = 0, no operations are needed to transform the first zero characters of S₁ into the first zero characters of S₂.
- D(i, 0) means transforming the first *i* characters of S_1 into the first **zero** characters of S_2 . How many operations?

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- Let's see if we can find some base conditions. D(0,0) = ?
- Surely, D(0,0) = 0, no operations are needed to transform the first zero characters of S₁ into the first zero characters of S₂.
- D(i, 0) means transforming the first i characters of S₁ into the first zero characters of S₂. How many operations?
 One needs to delete i characters, we therefore have,

$$D(i,0)=i$$

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- Let's see if we can find some base conditions. D(0,0) = ?
- Surely, D(0,0) = 0, no operations are needed to transform the first zero characters of S₁ into the first zero characters of S₂.
- D(i, 0) means transforming the first i characters of S₁ into the first zero characters of S₂. How many operations?
 One needs to delete i characters, we therefore have,

$$D(i,0)=i$$

Similarly, to transform the first *j* characters of S₂ into the first **zero** characters of S₁, i.e. D(0, *j*), we have delete the first *j* characters of S₂,

$$D(0,j)=j$$

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For the general case, how was D(i, j) obtained?



For the general case, how was D(i, j) obtained? Clearly, it was obtained by applying one of the three (four) possible edit operations: insertion, deletion, substitution (match), to a smaller alignment

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Given, two sequences S₁ and S₂ of length m and n respectively, which particular value of D solves the problem?

- For the general case, how was D(i, j) obtained? Clearly, it was obtained by applying one of the three (four) possible edit operations: insertion, deletion, substitution (match), to a smaller alignment
- Given, two sequences S₁ and S₂ of length m and n respectively, which particular value of D solves the problem?
- D(m, n) is the value that we are looking for. It is the minimum number of edit operations that are needed to transform the first *m* characters of S₁ into the first *n* characters of S₂

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 Recurrence
 equation
 for
 the
 edit
 distance
 problem

Base conditions,

$$D(0,0) = 0$$

 $D(i,0) = i, i \in 1...n$
 $D(0,j) = j, j \in 1...m$

General case,

$$D(i,j) = \min \begin{cases} D(i-1,j) + 1, \\ D(i,j-1) + 1, \\ D(i-1,j-1) + 1, \text{ if } S_1(i) \neq S_2(j), \\ D(i-1,j-1) + 0, \text{ if } S_1(i) = S_2(j), \end{cases}$$

Solution,

D(m, n)

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 Algorithm
 for solving the edit distance recurrence
 equation

Two strategies:

- Top-down
- Bottom-up

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 Top-down
 computation

In the top-down computation, a first call is made to compute D(m, n), which will force the computation of D(m − 1, n), D(m, n − 1) and D(m − 1, n − 1)

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- Many values will be re-computed several times in the top-down computation
- It is easy to see that an exponential number of operations will be performed!
- A complete 3-way tree of depth m has $\Theta(3^m)$ nodes.



Hum, but there are only $(n + 1) \times (m + 1)$ distinct D(i, j) values!



- Hum, but there are only $(n + 1) \times (m + 1)$ distinct D(i, j) values!
- The bottom-up computation proceeds with the small values of *i* and *j* first.



- Hum, but there are only $(n + 1) \times (m + 1)$ distinct D(i, j) values!
- The bottom-up computation proceeds with the small values of *i* and *j* first.
- Furthermore, the algorithm memorizes (caches) the values of D(i, j) so that a given D(i, j) is computed only once.



- 1. This technique is known as dynamic programming;
- 2. Dynamic programming can only be applied to problems with a structure known as the **Bellman principle**.

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Bottom up computation						

Bottom-up computation



where s(i,j) = 1 if $S_1(i) \neq S_2(j)$ and 0 otherwise.

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Preamble	Alignment	Dynamic programming	Dynamic programming	Traceback	References	
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 \Rightarrow Base conditions.

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 \Rightarrow Notice the two **alternatives**: D(1,2) + 1 = D(2,2) + 1 = 2

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 \Rightarrow The **final result** is D(4,5) = 2. What does it tell us?

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 \Rightarrow We now know that one sequence can be transformed into the other with as little as 2 edit operations!

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How do you fill up the matrix: by row? by column? by diagonal? it is not important?



- How do you fill up the matrix: by row? by column? by diagonal? it is not important?
- **How** many cells can be filled simultaneously? Leading to parallel computation.



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- ▶ D(4,5) = 2, it's possible to transform S_1 into S_2 with two edit operations, which ones?



- How do you fill up the matrix: by row? by column? by diagonal? it is not important?
- **How** many cells can be filled simultaneously? Leading to parallel computation.
- ▶ D(4,5) = 2, it's possible to transform S_1 into S_2 with two edit operations, which ones?
- **How** to compute the actual alignment?

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Dyna	mic Pro	ogramming				



 \Rightarrow How to recover the underlying alignment?

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Tracet	back				

In cell D(i, j):

set
$$\nwarrow$$
 if $D(i-1, j-1) + s(S_1(i), S_2(j)) = D(i, j)$,
set \leftarrow if $D(i, j-1) + 1 = D(i, j)$,
set \uparrow if $D(i-1, j) + 1 = D(i, j)$.

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To recover the edit transcript, the alignment, follow a path from D(n, m) to D(0, 0). Interpreting each pointer as follows:

- $\leftarrow: \text{ deletion of } S_1(j),$
- $\uparrow: \text{ insertion of } S_2(i),$
- \bigwedge : match of $S_1(i)$ and $S_2(j)$ if $S_1(i) = S_2(j)$ and substitution otherwise.

The two optimal alignments:

ATCGC ATCGC

A-GGC or AG-GC

 \Rightarrow It takes $\mathcal{O}(n+m)$ time to compute the traceback for one path.



There was more than one optimal alignment.

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- There was more than one optimal alignment.
- Only one solution was recovered, but we could have recorded all of them.

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- There was more than one optimal alignment.
- Only one solution was recovered, but we could have recorded all of them.
- How many optimal alignments are there?

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- There was more than one optimal alignment.
- Only one solution was recovered, but we could have recorded all of them.
- How many optimal alignments are there?
- Can you enumerate them?

Preamble	Alignment	Dynamic programming	Dynamic programming	Traceback	References
Refere	ences				

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Pensez-y!

L'impression de ces notes n'est probablement pas nécessaire!