Bioinformatics Algorithms

(Fundamental Algorithms, module 2)

Zsuzsanna Lipták

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Pairwise Alignment 3

Optimal pairwise alignment in linear space

Given two sequences s, t of length n:

- DP algorithm for global alignment: $O(n^2)$ time and space
- if we only want the score of an optimal alignment sim(s,t) (problem variant 1), then we can do this in $O(n^2)$ time and O(n) space (space-saving variant)
- But that algo does not give us the optimal alignment itself (problem variant 2)
- Now: algorithm for computing an optimal alignment itself in time $O(n^2)$ but space O(n)

There are several algorithms achieving this, e.g. Hirschberg (1975) aka Myers-Miller (1988). Here we present the divide-and-conquer algorithm from the book by Durbin, Eddy, Krogh, Mitchison (ch. 2.6).

$$s = GAAGA, t = CACA$$

match: 2, mismatch: -1, gap: -1

D(i,j)			C	A	С	A
		0	1	A 2	3	4
	0	0	-1	-2	-3	-4
G	1	-1	-1	-2	-3	-4
A	2	-2	-2	1	0	-1
A	3	-3	-3	1 0 -1	0	2
G	4	-4	-4	-1	-1	1
A	5	-5	-5	-2	-2	1

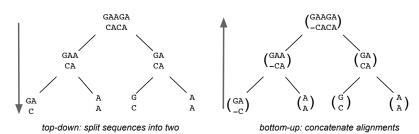
The optimal alignments are:

- 1. $\binom{\text{GAAGA}}{-\text{CACA}}$
- 2. (GAAGA
- 3. (GAAGA)
- 4. (GAAGA)

Consider the first optimal alignment (GAAGA):

Idea: Divide-and-conquer

We divide the two sequences s, t in two parts, left and right, align left with left, right with right, and then concatenate the two alignments:

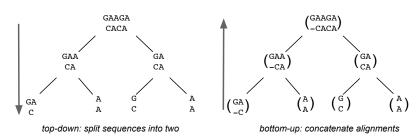


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Question

But how do we know where to divide them?

Definition

A cut is a pair of positions (n', m'), where $1 \le n' \le n$, and $1 \le m' \le m$ (with |s| = n, |t| = m).

- 1. In sequence 1, we will always take the middle cut position $n' = \lceil n/2 \rceil$.
- 2. In sequence 2, we will remember where the middle row $n' = \lceil n/2 \rceil$ was crossed.
- 3. For this, we will need to compute another matrix M (again, in space-saving manner!).

Matrix M

- Cell M(i,j) contains, where $i \ge n'$, an index r s.t. there is an optimal alignment with score D(i,j) passing through cell (n',r).
- Computation:

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M(n',j)=j for all j=1,\ldots,m; for i>n', 0\leq j\leq m: M(i,j)=M(i',j'), where D(i,j) derives from cell (i',j') (therefore (i',j')\in\{(i-1,j),(i-1,j-1),(i,j-1)\}) – if there is more than one, then choose one acc. to priority (e.g. left-diag-top)
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- Then M(n, m) = r s.t. there is an optimal alignment of s and t which passes through cell $(\lceil n/2 \rceil, r)$.
- Thus, we can use the cut $(n',r)=(\lceil n/2\rceil,M(n,m))$ in the divide-step and recurse with $s_1\ldots s_{n'}$ and $t_1\ldots t_r$ on the left, and $s_{n'+1}\ldots s_n$ and $t_{r+1}\ldots t_m$ on the right.

Algorithm PWA(s,t)

- 1. if $\max(|s|,|t|) \le 2$, then return an optimal alignment computed with N-W-algorithm
- 2. else
- 3. compute DP-table *D* row-wise, and
- 4. from $i = \lceil n/2 \rceil$ on, compute also matrix M (row-wise)
- 5. return $PWA(s_1 ... s_{\lceil n/2 \rceil}, t_1 ... t_r)$ concatenated with $PWA(s_{\lceil n/2 \rceil+1} ... s_n, t_{r+1} ... t_m)$.

(for a detailed example, see class notes)

Analysis

- Space: Since all matrix computations are row-wise, they all need linear space in m, and none need to be stored, thus O(m); we need to store the partial alignments, whose total length is the length of the final alignment, thus O(n+m): altogether space O(n+m)
- Time: In each iteration, we are exactly halving the problem size (wherever we cut t, string s is always cut in the middle), thus we get:

$$nm + \frac{1}{2}nm + \frac{1}{4}nm + \ldots \leq nm\sum_{k=1}^{\infty} \frac{1}{2^k} = 2nm \in O(nm).$$

Thus we doubled the time (asymptotically the same, both O(nm)), but reduced the space from quadratic to linear.