

A Local
Alignment
Approach to
RNA folding

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A Local Alignment Approach to RNA folding

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Background

Our Approach

Runtime
Analysis

Results

Conclusion

RNA Folding

RNA consists of the four base pairs Adenine (A), Guanine (G), Cytosine (C) and Uracil (U). These base pairs of RNA pair in a complementary fashion: Adenine to Uracil (A – U) and Cytosine to Guanine (C – G).

Unlike DNA for which we are concerned with optimally aligning two strands, for RNA we are concerned with how the strand folds with itself.

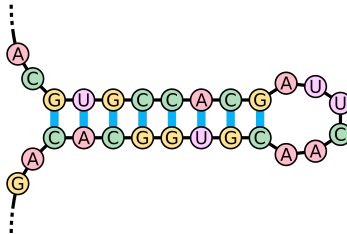


Figure: Source: <http://rosalind.info/problems/pmch/>

RNA Folding

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There are several frameworks with which we can model RNA folding. We will use the Energy Minimization Model.

In this model, matches are scored as $+1$ and non-matches as 0 . This lends itself to a dynamic programming algorithm: Nussinov's algorithm which we will call NUSSINOV .

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Let $r = r_1, \dots, r_n$ be a strand of RNA, where $r_i \in \{A, C, G, U\}$ and let $S(i, j)$ denote the optimal score of folding the subsequence $s_i, s_{i+1} \dots s_j \subset s$. Then,

$$S(i, j) = \max \begin{cases} S(i+1, j-1) + 1, & \text{if } i, j \text{ base pair} \\ S(i+1, j), \\ S(i, j-1), \\ \max_{i < k < j} \{S(i, k) + S(k+1, j)\}, & \text{bifurcation.} \end{cases}$$

It's clear that the runtime is $O(n^3)$.

k -Local Folding

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NUSSINOV returns the mathematically optimal alignment under the energy minimization model. Therefore, any new approach cannot hope to beat the scores, but only improve the running time.

Idea: Subsequences of an RNA strand which are (near) palindromes of each other are likely to be a good match. Pair these regions and pass the leftover segments to NUSSINOV .

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Formally, we define an algorithm `K-LOCAL FOLDING` which takes as input an RNA strand r and a parameter k , runs a local alignment algorithm on the strand to find k high scoring — and disjoint — palindromic regions of r . It then passes the remaining unpaired regions to `NUSSINOV` to be folded as usual.

K-LOCAL FOLDING

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```
1: procedure K-LOCAL FOLDING( $r, k$ )
2:   Initialize stack  $S \leftarrow r$ 
3:   while The number of local alignments found is  $\leq k$  and
       $S$  not empty do
4:      $s \leftarrow \text{pop}(S)$ 
5:     Let  $\bar{s}$  be the reverse of  $s$ 
6:     Call Local Alignment on  $s$  and  $\bar{s}$ .
7:     if Local alignment found then
8:       Remove the aligned regions from  $s$ 
9:       Push all unmatched regions of  $S$  back onto stack
10:    end if
11:  end while
12:  Call NUSSINOV on all unmatched regions of  $r$ .
13: end procedure
```

K-LOCAL FOLDING

example

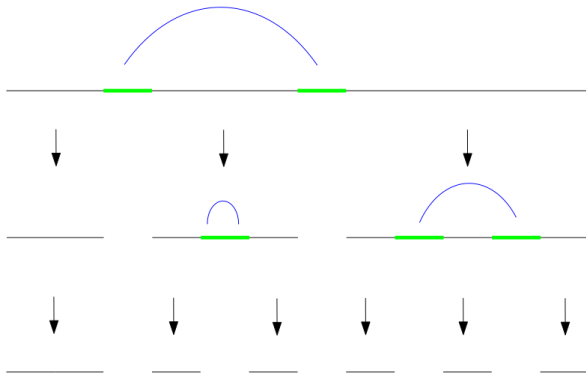


Figure: A strand of RNA undergoing multiple local alignments on successive subsequences. The unmatched regions at the bottom will be passed into NUSSINOV

k -Local Folding: Runtime Analysis

Fix an RNA strand r with length n .

Lemma

Let \mathcal{A} be the set of local alignments found. Finding k disjoint local alignments of r takes time

$$O\left(\left(n - \sum_{A \in \mathcal{A}} \ell_A\right)^2 \log k\right),$$

where ℓ_A is the length of an alignment $A \in \mathcal{A}$.

Lemma

K-LOCAL FOLDING *runs in time*

$$O\left(\left(n - \sum_{A \in \mathcal{A}} \ell_A\right)^3 + \left(n - \sum_{A \in \mathcal{A}} \ell_A\right)^2 \log k\right).$$

Proof of Lemma 1.

We can view the progress of the algorithm as a ternary tree: amortized across each level we perform local alignment on a sequence of size $n - \sum_{A \in \mathcal{A}} \ell_A$. Local alignment takes time squared in the size of the input. Since the tree has depth $\log(k)$, the result follows. ◀

Proof of Lemma 2.

Running NUSSINOV takes cubic time in the size of the input. K-LOCAL FOLDING first finds k disjoint alignments, then runs NUSSINOV on the remaining unmatched regions, which have total length $n - \sum_{A \in \mathcal{A}} \ell_A$. The result follows by applying lemma 1. ◀

k -Local Folding: Runtime Analysis

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We remark that since $\sum_{A \in \ell_A} \ell_A^3 \in O(n^3)$,
 $O((n - \sum_{A \in \mathcal{A}} \ell_A)^3) = O(n^3)$ so there is no asymptotic
difference between K-LOCAL FOLDING and NUSSINOV .

However, our hypothesis was that there may a difference in the
run times in practice.

Run Time Results

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Extensions and Further Research

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- 1 Modify K-LOCAL FOLDING to report possible pseudoknots
- 2 Compare K-LOCAL FOLDING to the Four Russians method of speeding up Nussinov.
- 3 Probabilistic Approach