Fast Local RNA Alignment

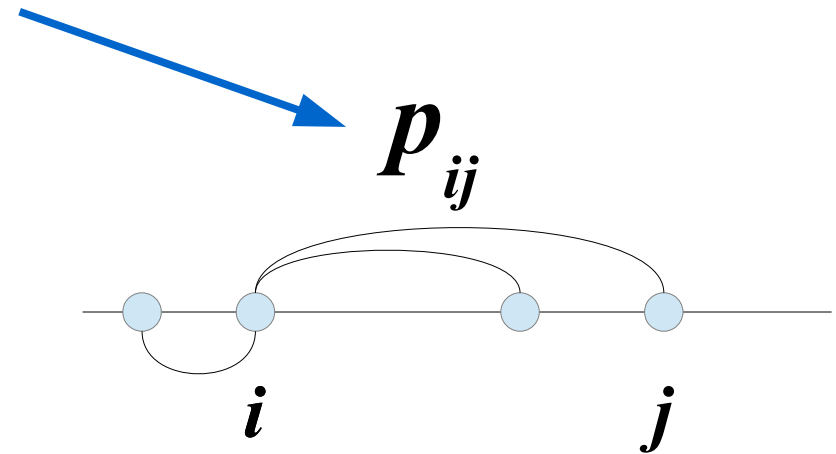Andrey Mironov
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Benasque 2015

Motivation

- Simultaneous alignment and structure prediction is very slow (Sankoff algorithm and its modifications)

- Most algorithms (especially for multiple alignment) produces global alignments: we need exact boundaries of the sequences

Idea1. Not to work with structure

- The bracket-dot presentation of the structure is a ***String***

- Analise the structure before the alignment

- Use the result of structure analysis in the alignment scoring
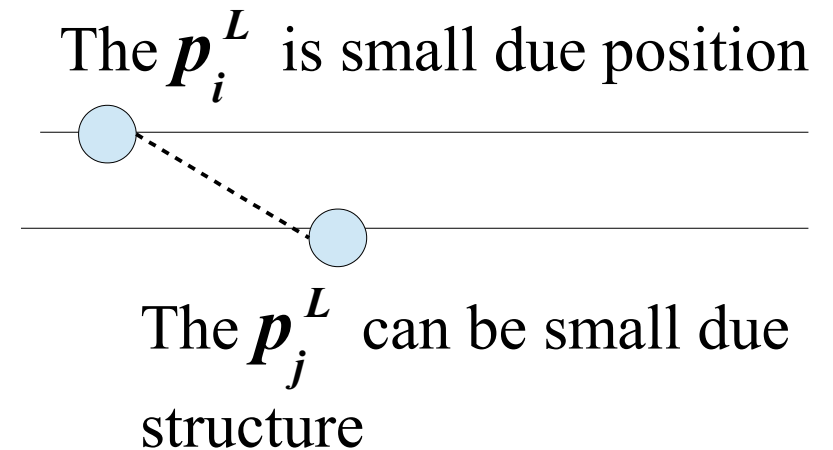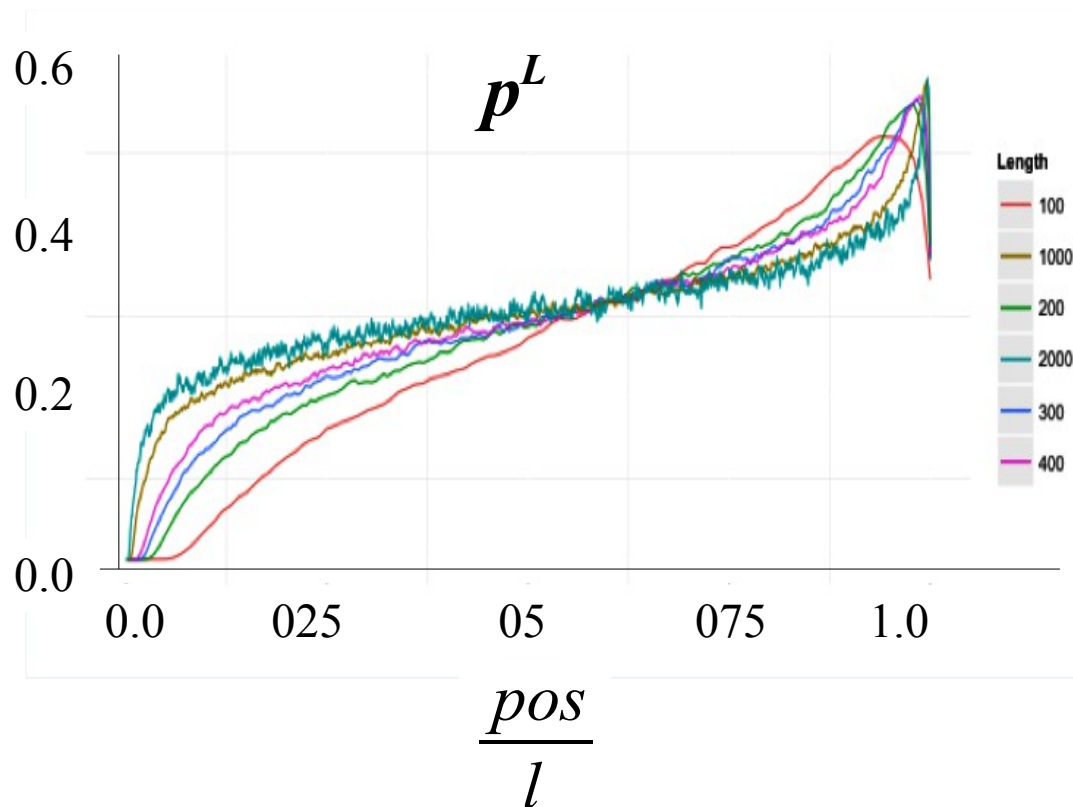 - Use the probabilities of pairing (pFold)

$$p_i^R = \sum_{j>i} p_{ij}$$

$$p_i^L = \sum_{j<i} p_{ji}$$

$$p_i^U = 1 - p_i^L \cdot p_i^R$$

$$\boldsymbol{p}_{ij}$$

$$i \qquad\qquad j$$

If $\{\boldsymbol{p^L, p^R}\}$ are similar the structures seems to be similar

The p^L and p^R can not be used directly

- The distributions of the p^L and p^R depend on sequence length and positions

The p_i^L is small due position

The p_j^L can be small due structure

The p_i^L, p_i^L are incomparable!

Rescaling of p^L, p^R

- We want to do a transformation of p^L, p^R to get a values with standard distributions that do not depend on the position and length.

- The *cdf* is uniformly distributed!

- The *cdf* can be fitted by:

$$cdf(x) = \alpha x^{b1} + (1-\alpha)\left(1-(1-x)^{b2}\right)$$

$$\alpha = \alpha(pos); \quad b2 = b2(pos); \quad b1 = b1(pos, l)$$

Scoring

$$W_{ij} = \alpha \cdot S_{ij}^{seq} + (1-\alpha) \cdot S_{ij}^{str}$$
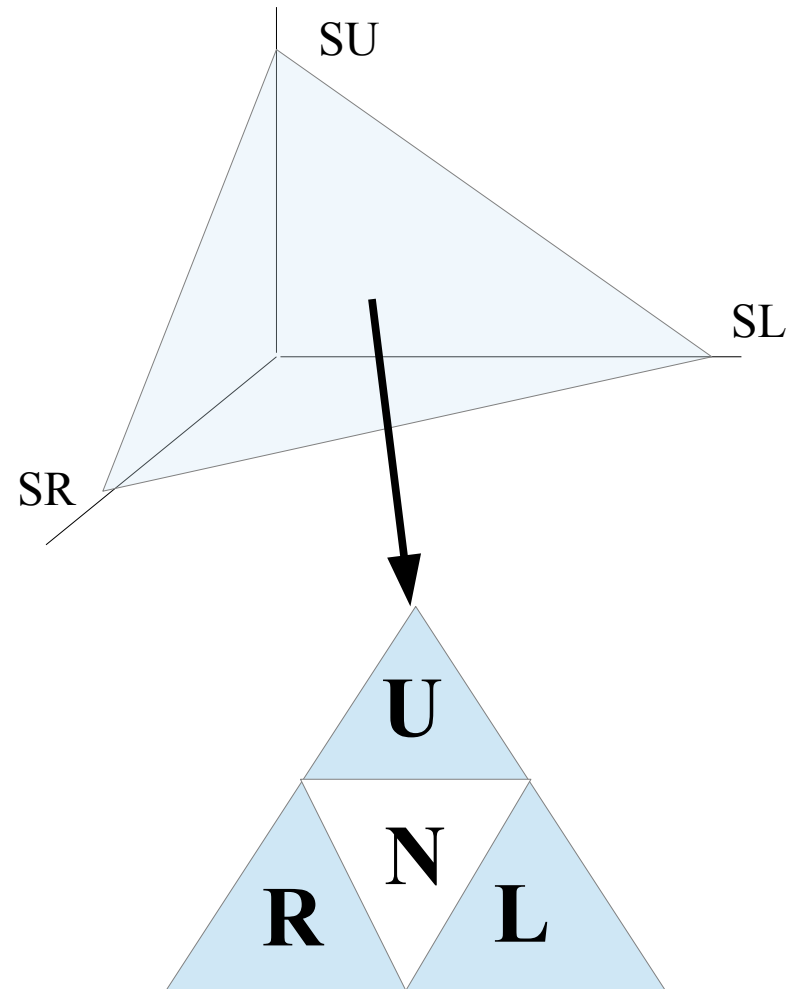$$S_{ij}^{str} = SL_{ij} + SR_{ij} + SU_{ij}$$

- S_{ij} are calculated as log-likelihood:

$$S_{ij}^{seq} = \log\left(\frac{p(s_i, s'_j)}{p(s_i)\,p(s'_j)}\right)$$
$$SL_{ij} = \log\left(\frac{p(s_i^L, s'^L_j)}{p(s_i^L)\,p(s^L{'}_j)}\right)$$
$$etc...$$

Idea 2. Non-progressive multiple alignment

- Do BLAST-like alignments between all sequences and find HSP

- Convert structure information to 4-letter alphabet

- Do BLAST-like alignments between all sequence structure signatures
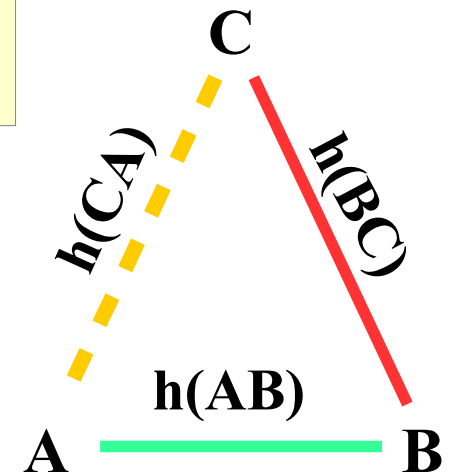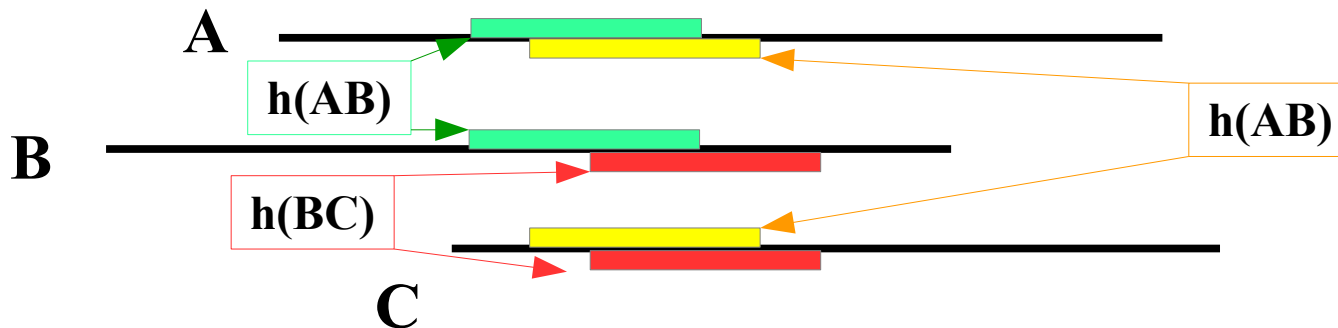
High Scoring Segments (HSP)

Definition1

$$HSP(A,B)=\{fA,tA;fB,tB\}$$
$$diag(HSP)=fA-fB$$

Definition2. *Two HSPs h(AB),h2(BC) for sequence pairs A~B and B~C are __compatible__ if there exist HSP h(CA) for pair C~A that:*
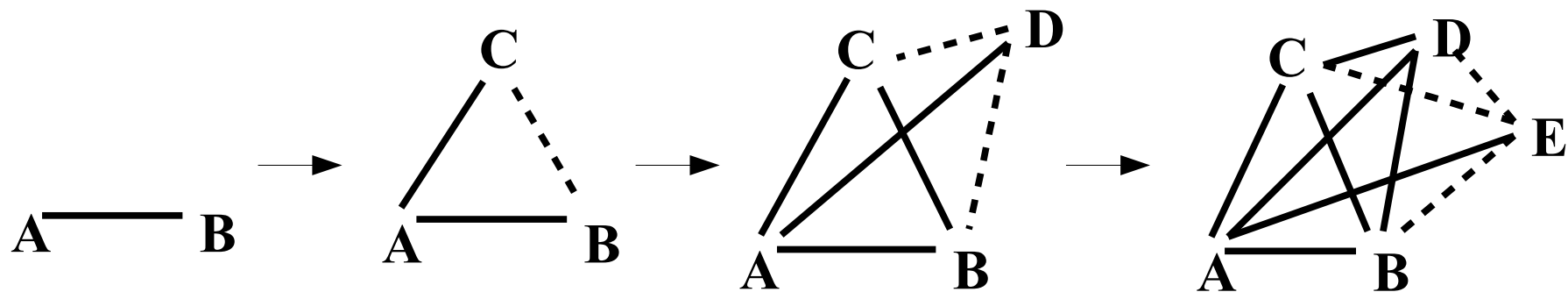
$$diag(h(AB))+diag(h(BC))+diag(h(CA))=0$$
$$iv(A,h(AB))\,overlaps\,iv(A,h(CA))$$
$$iv(B,h(AB))\,overlaps\,iv(B,h(BC))$$
$$iv(C,h(CA))\,overlaps\,iv(C,h(BC))$$
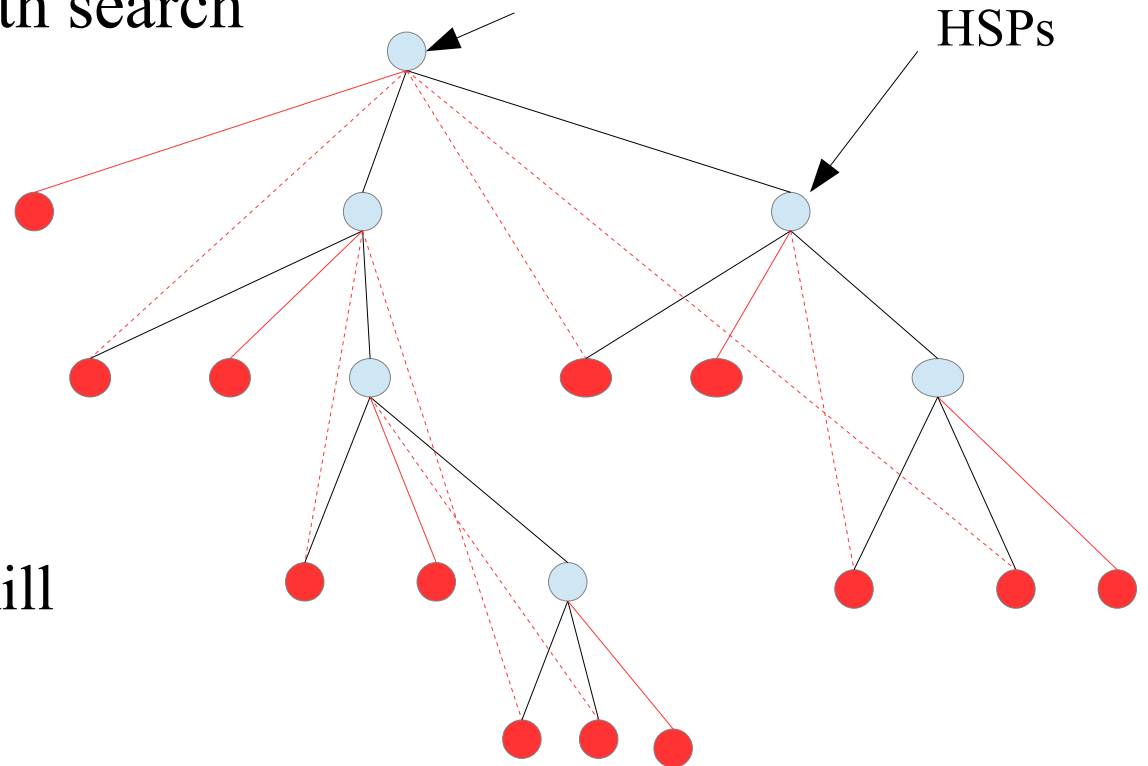
Search for sets of compatible HSPs (consensus set)

- Select a pair of the sequences (A,B)

- Select next HSP h1(A,B)

  - Select next HSP h2(AC) that is compatible with h1

    - Select next HSP h3(AD) that is compatible with h1 and h2

- ...

This is a clique problem (NP-hard), BUT...

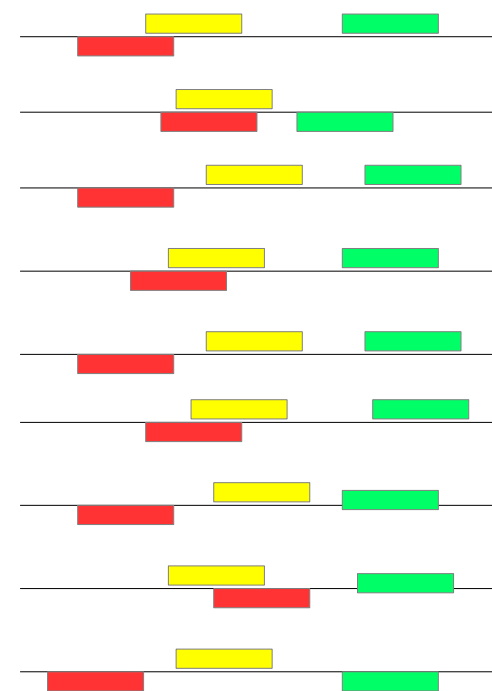The algorithm is in-depth search

HSPs

On every level we do more comparisons and have more chances to kill the recursion

Theoretically the expected number of iterations on a random sequences tends to a constant when number of sequences tends to infinity
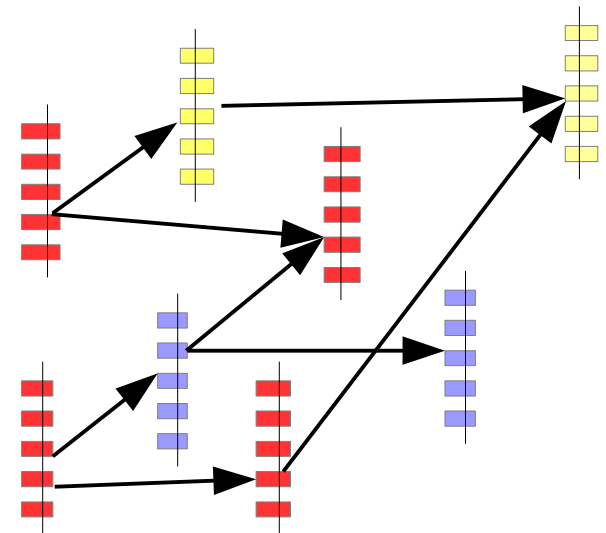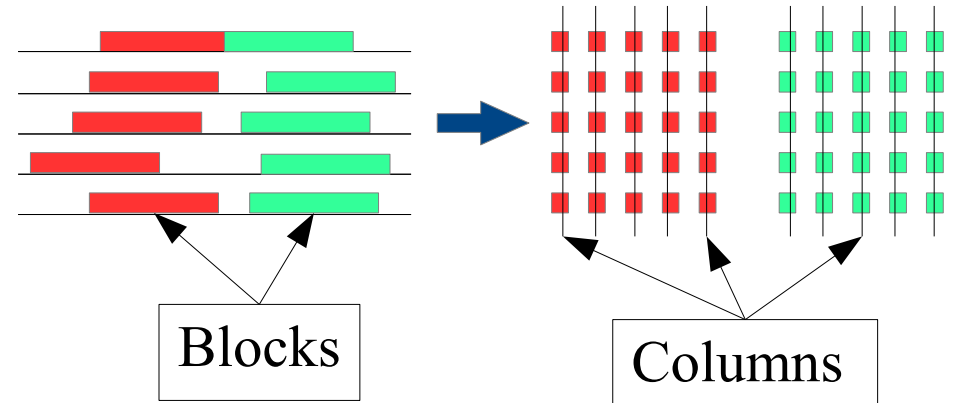
Algorithm: search for blocks

- Calculate p^L, p^R

- Transform structure information to structure alphabet

- Do BLAST-like search using sequences and structure

- Select combinations of HSPs that are compatible for all pairs of the sequences

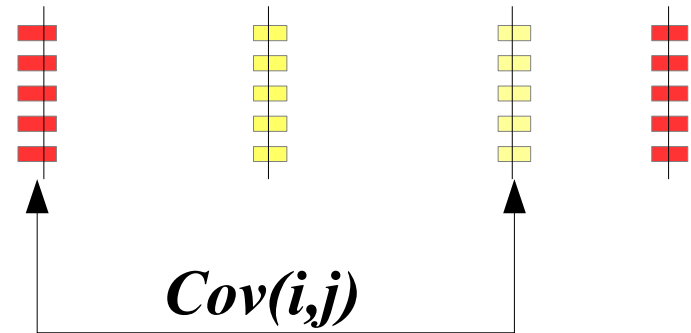- Search for consensus HS blocks (HSB)

Algorithm: alignment

- Decompose HSBs to a set of columns

- Column Graph (CG):

  - Vertices = columns

  - Edge $e=(u\rightarrow v)$ if for all sequences position i $v_i > u_i$; v_i, u_i position on sequence #i

- Do Dynamic Programming on CG and find the optimal alignment

Blocks

Columns

If you have enough time

- Calculate covariance between columns

$Cov(i,j)$

- Reconstruct optimal common structure and produce the alignment simultaneously (to be done)

Preliminary results

Without covariance

- tRNA with random flanks

- Identity 30-60%

- Quality (number of correctly aligned positions) = 80%

- Time for 20 sequences 2 s.

Variants

Variant 1

- Find HSS
- Near found diagonals do ProbCons-like alignment

Variant 2

- Do Nussinoff-style algorithm on columns

Variant 3

- Do hash-based partition function calculation

Team

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