# SPARSE: Quadratic Time SA\&F of RNAs without Sequence-Based Heuristics 

Sebastian Will
University of Leipzig

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## Simultaneous Alignment and Folding [Sankoff]

Given: $\begin{aligned} & A=\text { GCUGACGAGCACGCUCAUCGGUAAAUCUACCGAUCGUCAGCACU } \\ & \& \quad B\end{aligned}$

Find:

sequence similarity + energy $A+$ energy $B \rightarrow$ opt
where alignment, structure $A, \&$ structure $B$ are Compatible

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## Sankoff's SA\&F Algorithm

## Dynamic Programming

RNA Energy Minimization [Zuker]

$\otimes$
Sequence Alignment
$O\left(n^{6}\right)=$ "extreme computational cost"

## Sankoff-style Approaches <br> HEAVY

Dynalign FoldAlign

- Sankoff implementations - full ("heavy") energy model
- (sequence-based) heuristics


## PMcomp

- lightweight energy model
- base pair probabilities


## LocARNA

+ sparsifies structure space (ensemble-based)
- improves time and space


## RAF

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- strong sparsification w/o sequence-based heuristics


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## Sankoff: sequence similarity + energies of $A$ and $B \quad \rightarrow$ opt

- energy composed of loop energies
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Base Pair Maximization [Nussinov] \& Sequence Alignment

- cheaper but same complexity


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## LocARNA's Trick: <br> Ensemble-based Sparsification

- Sparsify structure ensemble

- improves time and space; each by $O\left(n^{2}\right)$


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all base pairs


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only probable base pairs
- improves time and space; each by $O\left(n^{2}\right)$


## SPARSE: Novel Ensemble-based Sparsification*



- only base pairs with probabilities $>\theta_{1}$
- only bases with unpaired probabilities in loops $>\theta_{2}$ - only base pairs with probabilities in loops $>\theta_{3}$ requires complete prediction (Sankoff/PARSE)
(*) confer LocARNA's "old" sparsification:
- match only base pairs with probabilities $>\theta_{1}$


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$w /$ complete: $a_{3}$ in loop $a_{2}$ $w / o$ complete: $a_{3}$ in loop $a_{1} X$

$$
a_{2} \boldsymbol{X} \Longrightarrow a_{3}-b_{2} \boldsymbol{X}
$$

## Thresholds in Recursions Cases



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## all base pairs $\theta_{1}$



## Modify Evaluation to Save Time



Quadratic Time


Q: How many matrices $M^{a b}$ compute $(i, k)$ ?


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A: each $(i, k)$ in only constant number of matrices

## (S)PARSE improves prediction over LocARNA

## LocARNA:

$\qquad$
A -CAACUCUGGAGAGUGUUUACGAAGGUAAACCACCCACGA
B UCGACCCUCGCGGGAGACAUCGGGAUU----CGAUCCCGA
$\qquad$
. (( $\ldots((((\ldots(((\ldots))) \ldots)))) .))$.
A AGCAAAUAUUUGUUCUUUUUUGAAGAAUGAAUAUGCAACU
B GGCCGA-AGGCGCAACCGCCCCGGAAACGCUCAGGCAA--$.(((\ldots-.((((\ldots(((\ldots))) \ldots)))) \ldots))) \ldots-$

## )) ). .

A UUCUGGUAUAAGGACAGAGAUUUCUUC
B ---------AAGGACCG----CGCGGG
---------. . . ...---- .))) . .

SPARSE:
 A ----CAA-CUCUGGAGAGUGUUUACGAAG-GUAAACCACC B UCGACCCUCGCGGGAGACAUCGGGAUUCGAUCCCGAGGCC

$\ldots((((\ldots(((()(((()((\ldots))))))))))))$. A CACGAAGCAAAUAUUUGUUCUUUUUUGAAGAAUGAAUAUG
B GAAGGCGCAACCG $\qquad$ CCC $\qquad$ CGGA
... ( ( ( . . . ( ( $\qquad$ - $\qquad$ )) ) .

> ...()) (..) )
$\qquad$ ))
A CAACUUUCUGGUAUAAGGACAGAGAUUUCUUC
B -AACGCUCAGGCAAAAGGACCGCGCGGG----
(..))))..) ) .......)) )) .) )


## Run times and speedup



Bralibase 2.1, pairwise alignments (k2)

Alignment Accuracy (Bb 2.1, k2)


## Structure Prediction Accuracy (BB 2.1, k2)



## Conclusions

## SPARSE: very efficient RNA alignment without sequence-based heuristics


http://www.bioinf.uni-freiburg.de/Software/SPARSE/

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- PARSE is THE lightweight Sankoff variant (cf. PMcomp)
- predicts deleted/inserted base pairs; like original SA\&F
- SPARSE $=$ Sparsified PARSE
- Novel ensemble-based sparsification (in-loop probabilities)
- No sequence-based heuristics
- Speeds up SA\&F: Quadratic Time $\left[\leftarrow O\left(n^{6}\right)\right]$
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## Thanks

...for your attention
... to my coauthors

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- Milad Miladi
- Mathias Möhl
- Rolf Backofen
... and the German Research Foundation $\boldsymbol{D} \boldsymbol{F}$

Appendix

## Computing "In Loop" Probabilities

from McCaskill matrices: $Q_{b}, Q_{m}$

similar: $\operatorname{Pr}[k$ unpaired in loop of $(\mathbf{i}, \mathbf{j})]$
[ExpARNA-P; Schmiedl et al., BMC Bioinformatics 2014]

## Alignment and Prediction Accuracy (Bralibase 2.1, 3-way alignments)



## SPARSE Improves Over LocARNA for Specific Families



(shown: IRES HCV, pairwise)

