

Curating and evaluating RNA structure assignments

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Outline:

- Part I: Semi-Automated RNA Sequence Editor (SARSE).
 - Curating RNA structural alignments.
 - Rnadbtools and SARSE.
 - Integrating Pfold and Pcluster.
 - The temperature of Rfam.
- Part II: R_K : The K category correlation coefficient.
 - Comparing two K category assignments.
 - Pearson's correlation coefficient and least square fitting.
 - Extending Pearson's correlation coefficient to two K dimensional tables the R_K coefficient.
 - Discretization, an extension of Matthews correlation coefficient.
 - Applications of R_K .

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Semi-Automated RNA Sequence Editor (SARSE)

Motivation:

- Good curated alignments to compare with predictions.
- Some years ago (still valid): Most ncRNA structural alignments have bad quality and obvious contain inconsistencies.
- Some databases even lack structural assignments corresponding to the multiple RNA alignment.
- Some years ago SRP RNA one of the best alignments, but had many inconsistencies. Old clean: from 20 to 3 pr sequence.
- Doing this kind of work is extremely useful to the community, but also extremely low prestige (and no funding :-(..).)
- Exist no good editor which include basic editing functions combined with structural consistency checks.

RNAdbtools: <http://rnadbtool.kvl.dk>

- Toolbox to conduct basic consistency checks.
- Highlights any tyupe of non-standard RNA pairs (and check, whether bases assigned to the same pair).
- Extends RNA stems where possible.
- Automated search and align0 realignment of global regions of blast hits. [Now outdated].
- Introduction of the column format: <http://colformat.kvl.dk>.
- Colformat motivation. Easy to work with while very flexible. [hence much man power in time are saved!]

We would like to have SARSE

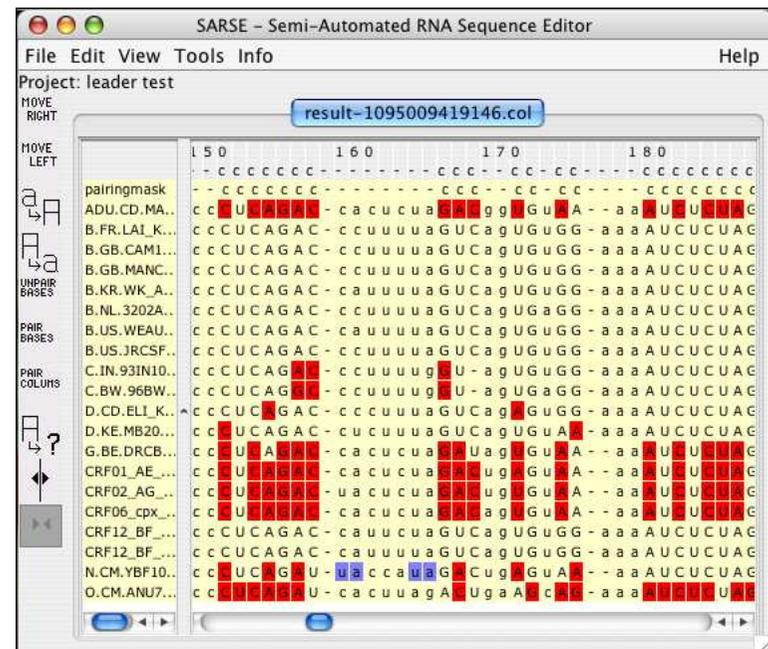
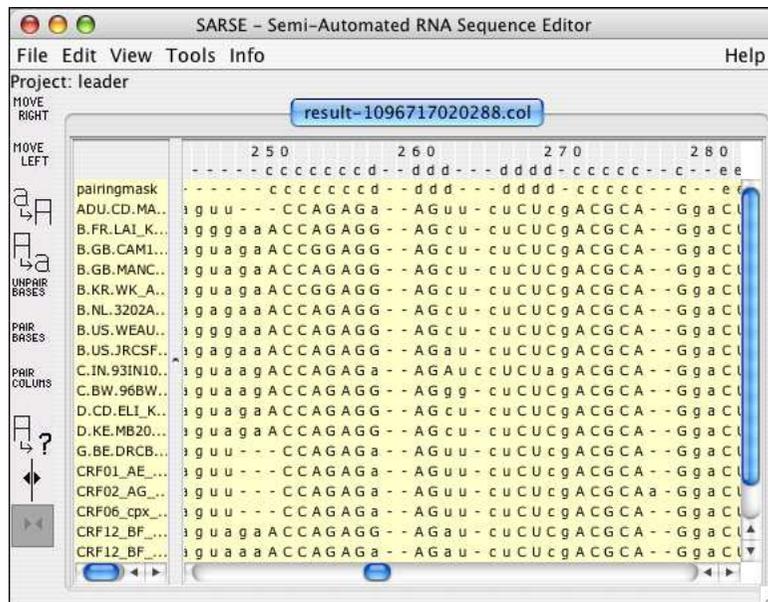
- Make RNAdbtools interactive.
- Features: Split view; drop drag; highlight complement bases; unlimited undo/redo sessions; overview (click and jump to region); history window.
- Integrate your own commandline software into SARSE [by dumping data in the colformat].
- Extends RNA stems where possible.
- Automated search and alignment of global regions of blast hits. [Now outdated, but similar stuff could be included in SARSE].
- SARSE: Java based interface. Basic editor functions directly incorporated.
- Other software can be executed: RNAdbtools, pfold and pcluster [NEW !].
- Extensive documentation: <http://sarse.kvl.dk>.

Semi-Automated RNA Sequence Editor

Clean up RNA multiple structural alignments. <http://sarse.kvl.dk> (See intro)

The screenshot displays the SARSE - Semi-Automated RNA Sequence Editor interface. The main window shows a multiple sequence alignment of HIV-1 RNA sequences. The alignment is displayed in a grid format with columns representing positions and rows representing individual sequences. The sequences are color-coded by reliability: green for double-stranded regions (>70% reliability) and blue for single-stranded regions (>70% reliability). The alignment is titled "result-1127982993814.col". Below the alignment, there are "Overview" and "History" panels. The "Overview" panel shows a bar chart of the alignment. The "History" panel lists recent actions such as "To Upper Case", "Move left", and "Move right". To the right of the alignment, there is a 3D RNA secondary structure diagram for sequence "B_HXB2_K03455". The diagram shows the RNA backbone with various structural elements labeled 1 through 9. A legend below the diagram indicates the reliability of the structure: green for double-stranded (>70% reliability) and blue for single-stranded (>70% reliability). At the bottom of the screenshot, a terminal window shows the command-line interface for the SARSE application, including the directory structure and the execution of the "pcluster" and "pfold" programs. The terminal output shows the directory structure and the execution of the "pcluster" and "pfold" programs. The terminal window is titled "ebbe@momo-130207-dellp4: /home/ebbe/Apps/sarse-1.15/tutorial-data - Shell - Konsole".

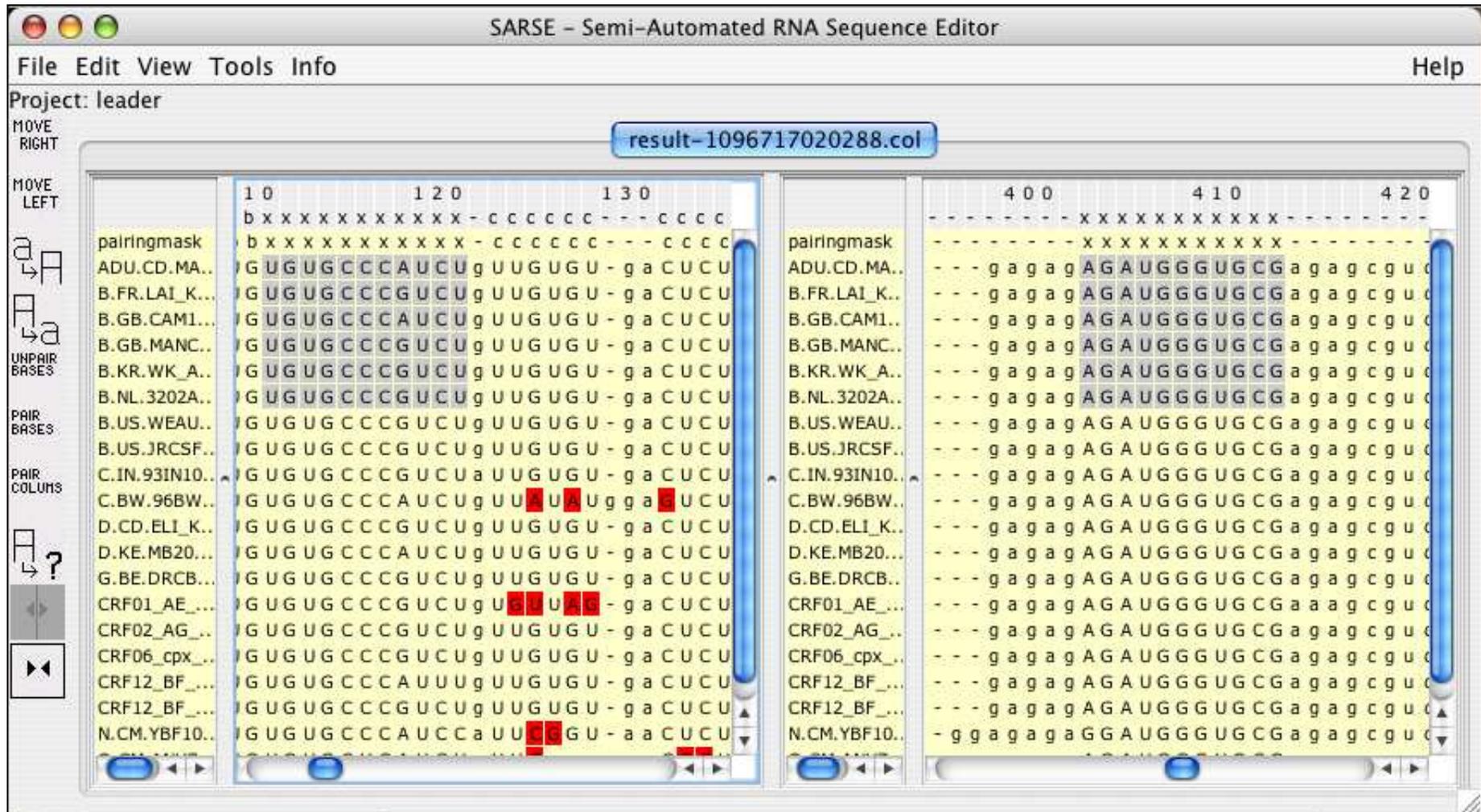
SARSE: Semi-Automated RNA Sequence Editor



- Invoke RNADBTOOLS to point out inconsistencies.
- Small overview box (not shown) gives a global view.
- Data from programs dumped in a projects directory.

SARSE: Semi-Automated RNA Sequence Editor

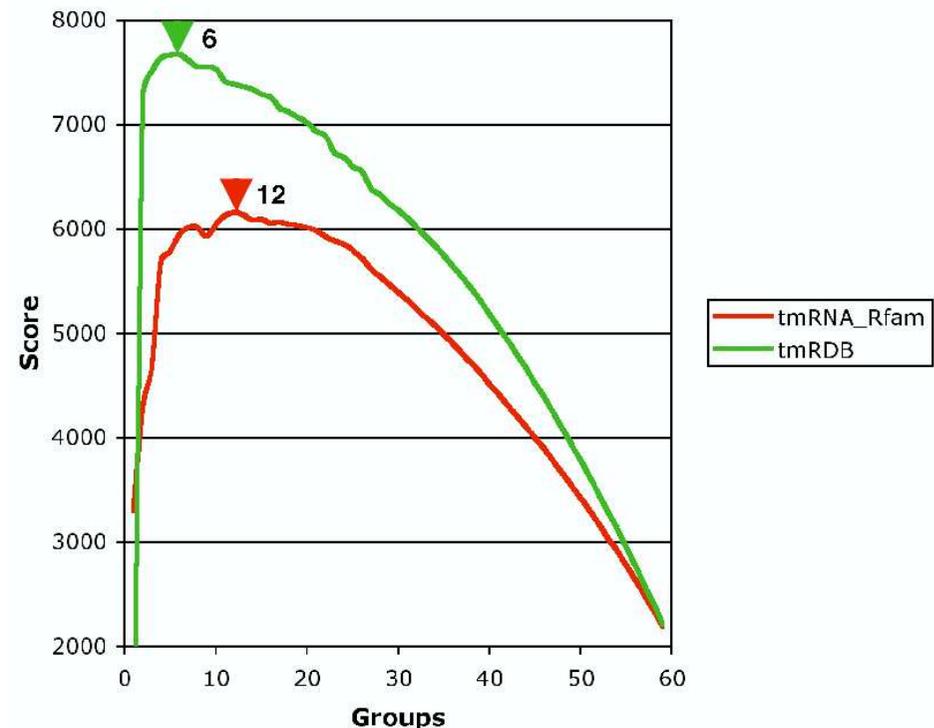
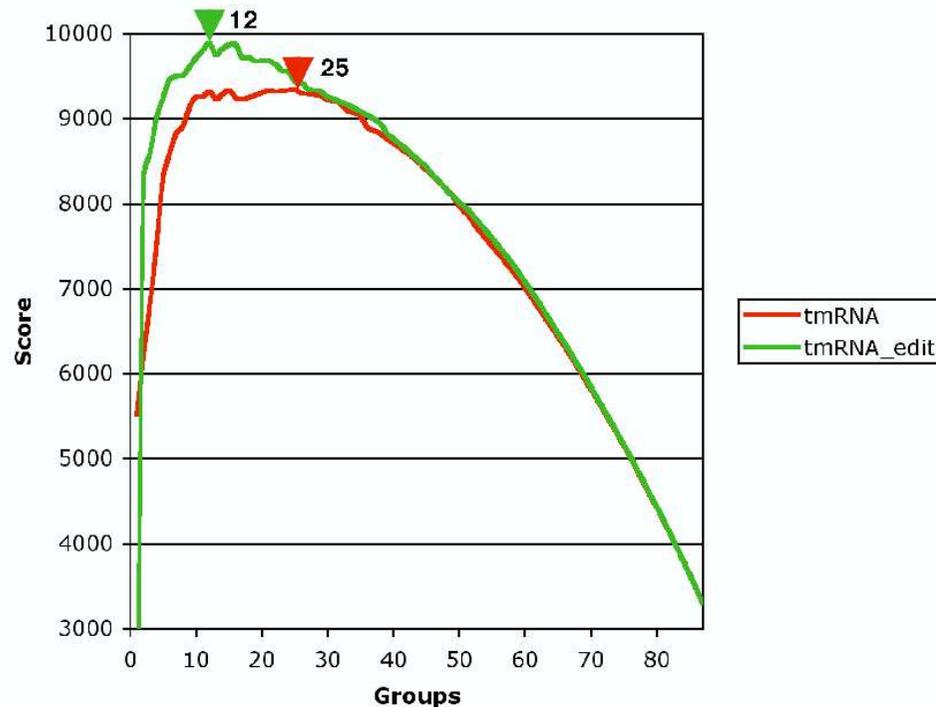
Split view



Semi-Automated RNA Sequence Editor: Pcluster

- pfold structure disrupted: poor alignment or variations in structure.
- Detect structural subgroups.
- Score: Reliability weighted sum of base pairs.
- Greedy heuristic method by joining subgroups with highest score.
- Extract: number of subgroups for max score.
- Heuristic to find "best" groupings by interpolation between max score point that score half of max point.

Semi-Automated RNA Sequence Editor: Pcluster



... and to something not so different:

You yeah I mean you: you train and test your prediction methods on these data

SARSE perspectives

- Adding even more programs. RNAalifold etc.
- Adding: auto fetching of sequences from databases.
- Web server set up.
- Web suites for specialist to curate their set of families.

Part II The K category correlation coefficient

Motivation:

- Predictions can yield more than dichotomies.
[Eg. protein secondary structure]
- RNA secondary structure predictions: bp, \neg bp, unassignable.
- Exists measure for comparing multiple categories
[eg: Escoufier (1973) and review by Baldi *et al.* (2000).]
- None of the measures have completely desired properties.
- Goal: an extension of Matthews correlation coefficient:

$$C = \frac{P_t N_t - P_f N_f}{\sqrt{(N_t + N_f)(N_t + P_f)(P_t + N_f)(P_t + P_f)}}$$

- Idea: Simple extension of Pearson's correlation coefficient.

Note on Matthews correlation coefficient for ncRNA evaluation

(... and to something *slightly* different!)

Basepair prediction N_t factor N larger than P_t, P_f, N_f . [$N(N-1)/2$ pairs of bases.]

$$\begin{aligned} C &= \frac{P_t N_t - P_f N_f}{N_t \sqrt{(1 + N_f/N_t)(1 + P_f/N_t)(P_t + N_f)(P_t + P_f)}} \\ &\approx \frac{P_t N_t - P_f N_f}{N_t \sqrt{(P_t + N_f)(P_t + P_f)}} \\ &= \frac{P_t}{\sqrt{(P_t + N_f)(P_t + P_f)}} \left[1 - \frac{P_f N_f}{P_t N_t} \right] \end{aligned}$$

where $N_f/N_t \rightarrow 0$ and $P_f/N_t \rightarrow 0$ for $N \rightarrow \infty$. For any reasonable prediction method ($P_t > 0$), with at least $P_t \sim P_f$ or $P_t \sim N_f$, we can write

$$C \approx \frac{P_t}{\sqrt{(P_t + N_f)(P_t + P_f)}} = \sqrt{\frac{P_t}{P_t + N_f} \frac{P_t}{P_t + P_f}},$$

Pearson's correlation coefficient and least square fitting

Pearson's correlation coefficient:

$$r = \frac{COV(X, Y)}{\sqrt{COV(X, X)COV(Y, Y)}}, \quad COV(X, Y) = \sum_{n=1}^N (X_n - \bar{X})(Y_n - \bar{Y})$$

For variables Y and X of length N least. Least square fitting in the coefficient b :

$$Y = a + bX$$

yield an expression for b . Conversely an similar expression can be obtained for fitting in the coefficient b' :

$$X = a' + b'Y$$

For a linear fit:

$$E = \sum_{n=1}^N (Y_n - (a + bX_n))^2$$

partial derivatives in a and b should be zero. It follows that

$$r^2 = bb', \quad b = \frac{COV(X, Y)}{COV(X, X)}$$

Extending Pearson's correlation coefficient to two K -dimensional tables, the R_K coefficient

Consider two $N \times K$ tables: $\underline{\underline{X}}$ and $\underline{\underline{Y}}$. Define

$$COV(\underline{\underline{X}}, \underline{\underline{Y}}) = \sum_{k=1}^K w_k COV(\underline{\underline{X}}_k, \underline{\underline{Y}}_k) = \frac{1}{K} \sum_{n=1}^N \sum_{k=1}^K (X_{nk} - \bar{X}_k)(Y_{nk} - \bar{Y}_k)$$

where $\bar{X}_k = \frac{1}{N} \sum_{n=1}^N X_{nk}$ and \bar{Y}_k are the respective means of column k . Use ("prior") $w_k = 1/K$.

$$R_K = \frac{COV(\underline{\underline{X}}, \underline{\underline{Y}})}{\sqrt{COV(\underline{\underline{X}}, \underline{\underline{X}})COV(\underline{\underline{Y}}, \underline{\underline{Y}})}}$$

Basic properties: $R_K \in [-1, 1]$; $R_1 = r$; $R_2 = r$, when $X_{n1} + X_{n2} = a$ and $Y_{n1} + Y_{n2} = b$. Hence R_2 reduces to Matthews correlation coefficient when X and Y components only take the values $\{0, 1\}$.

Relation to least square fitting

K related linear fits $\vec{Y} = \vec{a} + b\vec{X}$ over the N data points. $K = 1$: Pearson case.

Weighted difference in a cost function:

$$E = \sum_{n=1}^N \sum_{k=1}^K w_k (Y_{nk} - (a_k + bX_{nk}))^2$$

To obtain minimum. Require: $\partial E / \partial a_k = 0$ (for all $k = 1, \dots, K$) and $\partial E / \partial b = 0$.

After a little algebra:

$$\sum_{k=1}^K w_k \left(\sum_{n=1}^N X_{nk} Y_{nk} - N \bar{X}_k \bar{Y}_k \right) = b \left\{ \sum_{k=1}^K w_k \left(\sum_{n=1}^N X_{nk}^2 - N \bar{X}_k^2 \right) \right\}$$

yielding

$$b = \frac{COV(\underline{\underline{X}}, \underline{\underline{Y}})}{COV(\underline{\underline{X}}, \underline{\underline{X}})} \quad \text{and} \quad R_K^2 = bb'$$

The Discrete version of R_K

- The $K \times K$ confusion matrix $\underline{\underline{C}}$.
- Let C_{kl} be the number X_{nk} 's predicted to belong to class k , but belong to class l , $l \neq k$.
- For $K=2$: C_{11} : true positives; C_{22} : true negatives; C_{12} : false positives; C_{21} : false negatives.
- Well known observations:
 - $N = \sum_{kl} C_{kl}$.
 - $\bar{X}_k = \frac{1}{N} \sum_l C_{kl}$.
 - $\bar{Y}_k = \frac{1}{N} \sum_l C_{lk}$.
 - $C_{kk} = \sum_n X_{nk} Y_{nk}$.

The Discrete version of R_K

Plug in the known observations to R_K and obtain

$$R_K = \frac{\sum_{klm} C_{kk}C_{lm} - C_{kl}C_{mk}}{\sqrt{\sum_k \left(\sum_l C_{kl} \right) \left(\sum_{\substack{l' \\ k' \neq k}} C_{k'l'} \right)} \sqrt{\sum_k \left(\sum_l C_{lk} \right) \left(\sum_{\substack{l' \\ k' \neq k}} C_{l'k'} \right)}}$$

or equivalently

$$R_K = \frac{N \text{Tr}(\underline{\underline{C}}) - \sum_{kl} \underline{\underline{C}}_k \hat{\underline{\underline{C}}}_l}{\sqrt{N^2 - \sum_{kl} \underline{\underline{C}}_k (\hat{\underline{\underline{C}}}_l^\top)_l} \sqrt{N^2 - \sum_{kl} (\underline{\underline{C}}_k^\top)_k \hat{\underline{\underline{C}}}_l}}$$

- $\underline{\underline{C}}_k$ the k th row of $\underline{\underline{C}}$.
- $\hat{\underline{\underline{C}}}_l$ the l th column of $\underline{\underline{C}}$.
- $\underline{\underline{C}}^\top$ is $\underline{\underline{C}}$ transposed.

Applications of R_K

Comparison to other measures of evaluating protein secondary structure predictions [From EVA (Rost and Co-workers)]

- Numerous approaches for protein secondary structure prediction.
- Predicting the three classes, α -helix, β -sheet and coil.
- Comparing to Q_3 ranking, the fraction of correctly predictions over all three classes.
- Comparing to SOV (Segment Overlap), measure that take continuous stretches of helices and sheet into consideration in the evaluation.

Applications of R_K

- Eva (as of August 2003) have several classes of different set sizes.
- Each set covers different number of predictions methods.

R	1	0	0	1	0	0
L	1	0	0	1	0	0
R	1	0	0	1	0	0
V	1	0	0	1	0	0
H	1	0	0	1	0	0
Q	1	0	0	1	0	0
I	1	0	0	1	0	0
A	1	0	0	1	0	0
E	1	0	0	1	0	0
E	1	0	0	1	0	0
H	0	0	1	1	0	0
G	0	0	1	0	0	1
L	0	1	0	0	0	1
R	0	1	0	0	0	1
H	0	1	0	0	0	1
D	0	1	0	0	0	1
S	0	1	0	0	0	1
S	0	1	0	0	0	1
G	0	0	1	0	0	1
G	0	0	1	0	0	1
E	0	0	1	0	0	1
K	0	0	1	0	0	1
	α	β	C	α	β	C

Set	method	R_3	rank	sov	Q_3
1	profsec	0.621	1	74.8	75.09
	psipred	0.619	1	73.6	75.11
	apssp2	0.613	1	71.4	74.74
	samt99_sec	0.613	1	71.1	74.68
	sspro2	0.598	1	69.1	73.81
	phdpsi	0.581	1	69.7	72.61
	jpred	0.570	1	70.3	71.81
	prospect	0.567	1	69.8	71.77
	prof_king	0.555	1	69.8	70.83
phd	0.526	2	64.5	69.01	
2	profsec	0.600	1	71.5	74.00
	samt99_sec	0.586	1	67.1	73.28
	psipred	0.579	1	69.8	72.86
	sspro2	0.573	1	67.4	72.50
	phdpsi	0.560	1	66.9	71.62
	prof_king	0.544	1	66.5	70.33
	jpred	0.536	1	66.7	69.92
phd	0.505	2	62.6	68.03	
3	profsec	0.600	1	71.5	74.00
	psipred	0.579	1	69.8	72.86
	samt99_sec	0.586	1	67.1	73.28
	phdpsi	0.560	1	66.9	71.62
	prof_king	0.544	1	66.5	70.33
	jpred	0.536	1	66.7	69.92
phd	0.505	2	62.6	68.03	
4	profsec	0.608	1	71.7	74.61
	psipred	0.591	1	71.1	73.69
	samt99_sec	0.591	1	68.9	73.69
	phdpsi	0.568	1	67.7	72.22
	jpred	0.545	1	67.5	70.63
phd	0.512	2	64.5	68.54	
5	psipred	0.608	1	71.0	74.82
	profsec	0.606	1	70.2	74.52
	samt99_sec	0.600	1	69.5	74.32
	phdpsi	0.566	2	66.6	72.13
	phd	0.533	3	64.8	69.95
6	psipred	0.617	1	72.1	75.44
	profsec	0.610	1	71.2	74.82
	phdpsi	0.565	2	67.5	72.10
	phd	0.540	3	65.9	70.43

Applications of R_K

RNA example of R_3 . First some background:

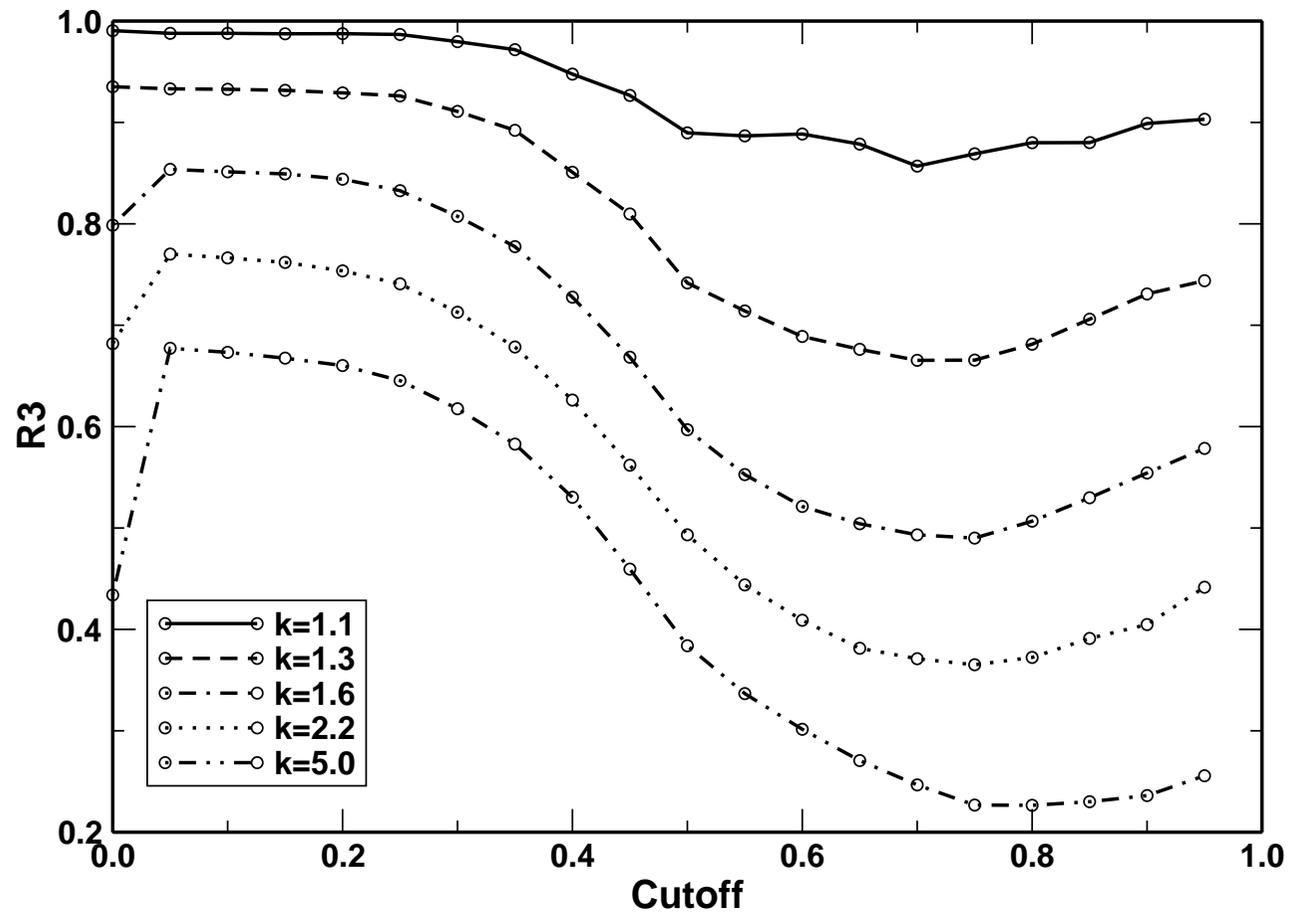
- Study applying pfold to structure prediction of the HIV leader (713 nt.) using 20 aligned sequences (Knudsen *et al.*, 2004).
- Study: Compare pfold predictions to predictions with perturbed rate values.
- Aim of study: can the evolutionary rates estimated from rRNA and tRNA be applied on the much faster evolving HIV-1 sequence.
[NOTE: this assumption is implicit for all prediction methods estimating parameters from *e.g.*, rRNA and tRNA, such as for QRNA.]
- The answer was yes; the HIV-1 prediction is fine, just within the limits stabil predictions obtained when perturbing the evolutionary rates.
- Rates were essentially perturbed by having 50% chance of making the rate k times larger or k times smaller [and adding some normalization constraints].

Applications of R_K

- Introducing a third category, the unknown or unassignable categories.
- pfold predictions uncertain for low reliability score.
- Statement: If pfold score low for positions that are manually hard to assign basepairs, the overall prediction should be higher than if an assignment was enforced on these positions.
- Use different reliability score cut-offs for sending a basepair in the third category "unknown". The reference structure is still the original pfold prediction.

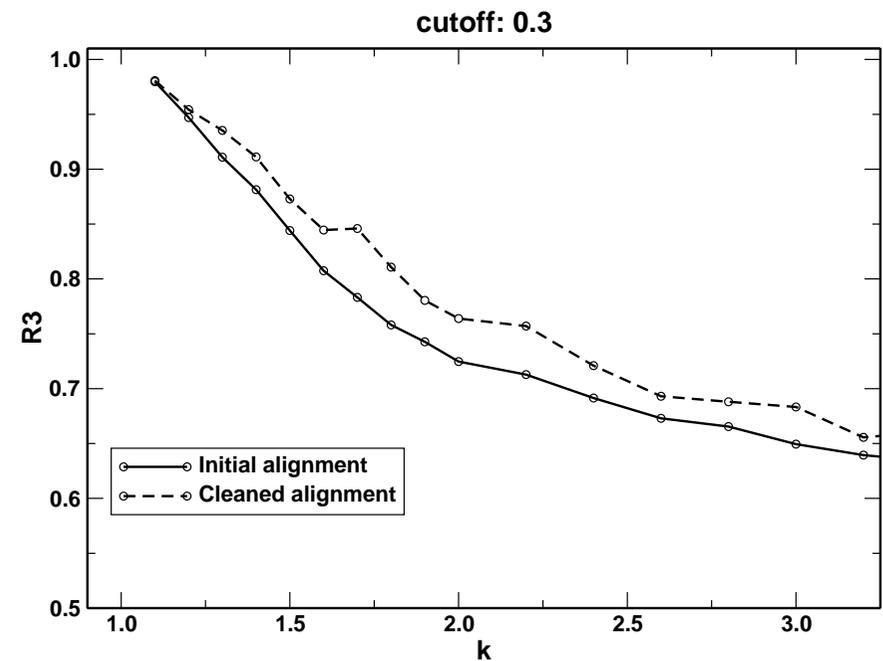
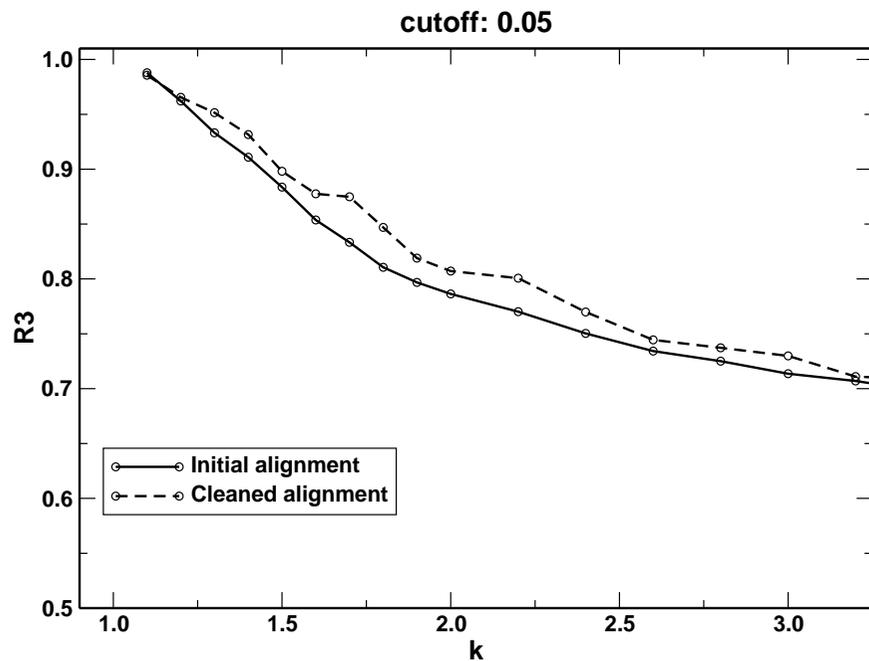
Applications of R_K

Rate variation performance for various cut-offs.



Applications of R_K

RNA structure predictions for different reliability (pfold) cut-offs for varying rate perturbations k . A comparison between initial and cleaned alignment.



Perspectives for R_K

Measures for comparing a predicted structure assignment to a curated structure assignment.

- Applying R_K to cases of $K > 3$.
- A measure as SOV would be needed to take prediction of entire helices into consideration.
- Further extension: Comparing $L \times N \times K$ tables and compute one correlation coefficient.