EECS730: Introduction to Bioinformatics

Lecture 11: Non-coding RNA discovery



http://desktop.jncasr.ac.in/uploaded/mrsrao/Slide1.jpg

Slides adapted from Dr. Shaojie Zhang (University of Central Florida)

Problem: how can we predict noncoding RNA genes from the genome

- We know that we can do this for protein-coding genes (gene-finding)
- Using HMMs that summarize the gene features
- However, noncoding RNAs are in general harder to detect
- No codon preference information available

Stable secondary structure?

• The stability of ncRNA secondary structure is not sufficiently different from the predicted stability of a random sequence. [Rivas and Eddy *Bioinformatics* (2000)]. ** CG rich random sequences ** 100 nts long **



RNA folding

- Algorithms/programs to compute the minimum energy:
 - Nussinov et al (1978), Waterman (1978), Smith and Waterman (1978), and Zuker and Sankoff (1984).
 - Mfold (Zuker 2003) and RNAfold (ViennaRNA) (Hofacker 2003).
- RNA folding via energy minimization has its shortcomings:
 - Prediction depends on correct energy parameters.
 - Sometimes, the true structure does not have the minimum energy.





Other information to use?

 Covarying mutations found from the multiple sequence alignment is a strong indication of RNA secondary structure



GCAUCG	ug	GUUC	aguggua	GAAU	//	CCGAUGCa
UCUAAUA	ug	GCAU	auua	GUGC	//	UAUUAGAa
GGGGAUG	ua	GCUU	aguggua	GAGC	//	UAUCCCCa
GCCGCCG	ua	GCUC	agcccggga	GAGC	//	CGGCGGCa
GGGCCCG	Jua	GCUU	agcucggua	GAGC	//	CGGGCCCa

Incorporating covarying mutation information

- If we have correct multiple alignments, looking for covarying mutations and finding consensus structure is a good way to do structure prediction.
 - RNAalifold (Hofacker et al. 2002)
 - The consensus structure prediction is more accurate.
 - To find energetically stable consensus structure is more statistically significant.
 - Still compute the MFE.
 - Covariance information is incorporated into the energy model by rewarding compensatory and consistent mutations.

$$E_{i,j} = \min \left\{ E_{i,j-1}; \min_{\substack{k: \ i+m < k \le j \\ \Pi_{ik} = 1}} E_{i+1,k-1} + E_{k+1,j} + \beta_{ik} \right\}$$

Incorporating covarying mutation

• Take into account covariance contribution:

$$d_{ij}^{\alpha,\beta} = 2 - \delta\left(a_i^{\alpha}, a_i^{\beta}\right) - \delta\left(a_j^{\alpha}, a_j^{\beta}\right)$$

$$C_{ij} = \frac{1}{\binom{N}{2}} \sum_{\alpha < \beta} d_{ij}^{\alpha,\beta} \Pi_{ij}^{\alpha} \Pi_{ij}^{\beta}$$

- Take into account inconsistent sequences: $q_{ij} = 1 \frac{1}{N} \sum_{\alpha} \left\{ \prod_{ij}^{\alpha} + \delta(a_i^{\alpha}, \text{gap}) \delta(a_j^{\alpha}, \text{gap}) \right\}$ where $\delta(a', a'') = 1$, if a' = a'' and 0, otherwise.
- Put together:

 $B_{ij}=C_{ij}-\phi_1q_{ij}$

 $\Pi_{ij} = 1$ if sequence positions *i* and *j* can form a base-pair, i.e. if (x_i, x_j) is in the set of allowed base-pairs $B = \{GC, CG, AU, UA, GU, UG\}$, and $\Pi_{ii} = 0$ if x_i and x_i cannot pair.

Mountain plot of 16s rRNA



De novo detection of RNA elements

- To find energetically stable consensus structure is more statistically significant.
- MFE can be used to compute the statistical significance.
 - MFE: *m*
 - Mean: *ų*
 - Standard deviation: δ
 - Z-score: *z* = (*m ų*)/δ
- We need randomize the multiple sequence alignment
 - Shuffle the columns of the input alignment
 - Not destroy the gap structure.
 - Certain sequence pattern.



Consensus Folding of Aligned Sequences as a New Measure for the Detection of Functional RNAs by Comparative Genomics

Stefan Washietl and Ivo L. Hofacker*



Figure 1. Mean z-scores of various RNA types dependent on the number of sequences in alignment. N=1 means RNAfold predictions for single sequences. Mean pairwise identities of the alignments are between 65% and 85%. See Table 1 for more details.



AlifoldZ

Real data performance

- Use MultiPipMaker to generate the multiple alignment of *S. cerevisiae* and other 6 related yeast genome.
- Extracted the regions of annotated ncRNAs
- Refine the poorly aligned regions
- Window size = 150, slide 20.
- False-positive rate: 0.25%.
- 30 CPU days.

ncRNA type	Annotated genes	Detected genes ($z < -4$)	Sensitivity (%)	
tRNA	275	28	10.2	
rRNA	11	6	55.5	
snRNA	6	4	66.7	
C/D snoRNA	46	5	10.9	
H/ACA snoRNA	20	14	70.0	
Other ncRNAs of known function	4	4	100.0	
ncRNAs of unknown function (RUF)	5	5	100.0	

Problem remains



- We need good multiple alignments to correctly predict secondary structures.
- We need to know the correct secondary structures to generate good multiple alignments.
- Solution:
 - Use Simultaneous Alignment and Folding (Sankoff Algorithm); computational intensive
 - Only apply on RNA sequences who have the "right" sequence similarity (between 60-95%)

RNAz (PNAS, 2005)

- z-score (for individual sequence)
 - Using Support Vector Machine (SVM) regression.
 - Using >10,000 point to define the independent variables (4-variables in total).
 - different length.
 - different base composition (GC/AT, A/T, G/C ratio).
 - Compute Mean (μ) and standard deviation (δ) for each data point to define the dependent variable
 - Compute the MFE of the sequence, and compute Z-score: $z = (m q) / \delta$
 - For an alignment, using the mean of the z-scores.

Z-score estimation



RNAz (classifying true/false noncoding RNA)

- Estimate a probability (P) if the alignment is classified as a structured RNA, based on
 - SCI
 - z-score
 - Average pairwise identity
 - Number of sequences.
- It is also done by SVM.

SCI (structure conservation index)

A much more efficient normalization can be achieved, however, by comparing the consensus MFE with the MFEs of each individual sequence in the alignment. To this end, we folded the alignment and calculated the consensus MFE E_A of the alignment by using RNAALIFOLD. If the sequences in the alignment fold into a conserved common structure, the average \overline{E} of the individual MFEs will be close to the MFE of the alignment, $E_{\rm A} \approx \overline{E}$. Otherwise, the MFE of the alignment will be much higher (indicating a less stable structure) than the average of the individual sequences, $E_A \gg \overline{E}$. We therefore define the SCI as

$$SCI = E_A / \bar{E}.$$

Classification based on z scores and SCI by a SVM



Structure conservation index



Performance of RNAz

Table 2. Detection performance (sensitivity/specificity) for SRP RNA and RNAseP alignments with mean pairwise identities between 60% and 90%

Program	2	3	10
QRNA	42.9/92.9		
DDBRNA	45.4/98.5	58.0/94.5	
MSARI			pprox 56/100
RNAZ	87.8/99.5	94.1/99.6	100/100

No. of sequences in alignment

Using RNAz to scan the human genome

- Nature Biotechnology 23, 1383 1390 (Nov. 2005), "Mapping of conserved RNA secondary structures predicts thousands of functional noncoding RNAs in the human genome"
- Input:
 - Genome-wide alignments of vertebrates from UCSC genome browser.
 - Using PhastCons program to find the most conserved
 - Adjacent conserved regions (<50 distances) are joined together.
 - All regions > 50 bps.
 - Remove all "known genes" and "Refseq genes"
- Output:
 - Predicted structured RNA elements in the human genomes using RNAz

Results

Table 1. Genomic coverage of filtering steps and phylogenetic conservation of ncRNA candidates.

	Genome Coverage		Alignments	RNAz hits $p > 0.9$		0.9
	Size	Fraction	Number	Size	Fraction of	Number
	(MB)	(%)		(MB)	input (%)	
Human genome	3,095.02	100.00	—			
PhastCons most conserved	137.85	4.81	1,601,903			
without coding regions	110.04	3.84	1,291,385			
without alignments $< 50nt$	103.83	3.33	564,455			
Set 1: 4 Mammals	82.64	2.88	438,788	5.46	6.62	35,985
Set 2: + Chicken	24.00	0.85	104,266	1.34	5.50	8,802
Set 3: + Fugu or zebrafish	6.86	0.24	30,896	0.14	2.03	996





p>0.9

p>0.5

0 -





