RNA structure determination

Experimental techniques

&

Computational prediction

Why study RNA structure?

Biological function highly depends upon RNA folding: The structure of an RNA molecule determines both the function of the molecule and the mechanism behind that function.

From Felden 2007:

"Proper functioning of RNAs requires the formation of intricate three-dimensional (3D) structures, as well as the ability to efficiently interconvert between multiple functional states."

RNA structure might reveal RNA's role in the origin and evolution of life on earth

RNA structure might function as drug target. Example: stem-loop II motif in RNA element of SARS virus genome (M.P. Robertson 2005)

Functions of RNA include:

- -- coding
- -- information transfer
- -- catalytic activities

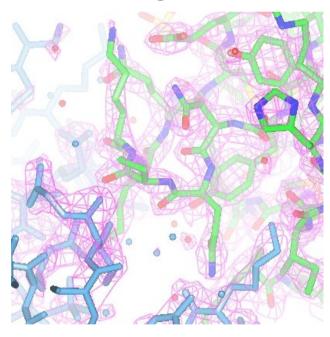
Structure determination Experimental techniques

B. Felden. Curr. Opinion in Microbiology (2007). 10:286-291

High resolution methods

X-ray crystallography

Nuclear Magnetic Resonance (NMR) spectroscopy



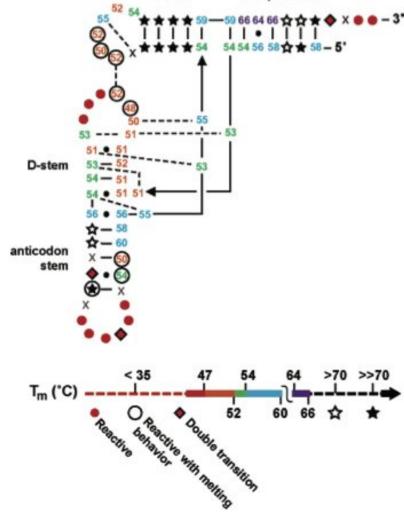


Cryo-electron microscopy (Cryo-EM)

Low(er) resolution methods

- -- Chemical/enzymatic probing
- -- Thermal denaturation (melting studies)
- -- Mass spectrometry
- -- RNA engineering

Selective 2'-Hydroxyl Acylation analyzed by Primer Extension (SHAPE chemistry)



Structure determination Computational structure prediction

Review articles:

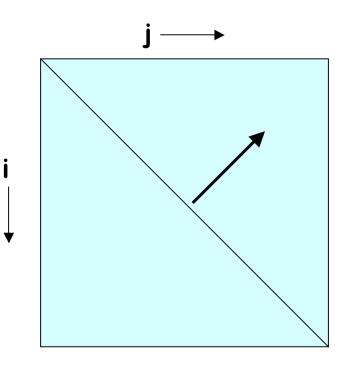
- -- M. Zuker. Curr. Opin. in Structural Biology (2000). 10:303-310.
- -- F. Major & R. Griffey. Curr. Opin. in Structural Biology (2001). 11:282-286.
- -- P.P. Gardner & R. Giegerich. BMC Bioinformatics (2004). 5:140.
- -- D.H. Mathews. J. Mol. Biol. (2006). 359:526-532.
- -- D.H. Mathews & D.H. Turner. Curr. Opin. in Struct. Biol. (2006). 16:270-278.
- -- Y. Ding. *RNA* (2006). 12:323-331.

Base pair maximization

Nussinov & Jakobson 1980

Dynamic programming

- -- fill stage
- -- traceback



Recursion rules

$$F(i,j) = max \begin{cases} s(x_i, x_j) + F(i+1, j-1) \\ F(i,k) + F(k+1, j) \end{cases} i \le k < j$$

Initialize F(i,i-1) = 0 for i = 2 to L

Base pair score F(i,i) = 0 for i = 1 to L s(xi,xj) = 1 if i and j base pair, 0 otherwise

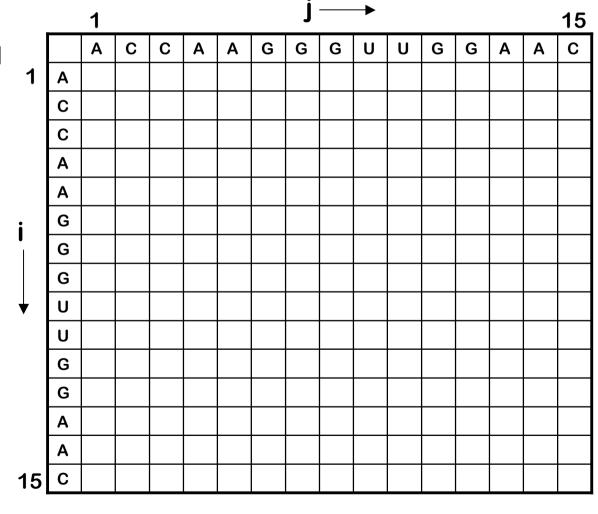
Nussinov example

Calculate the optimal folding for sequence: ACCAAGGGUUGGAAC

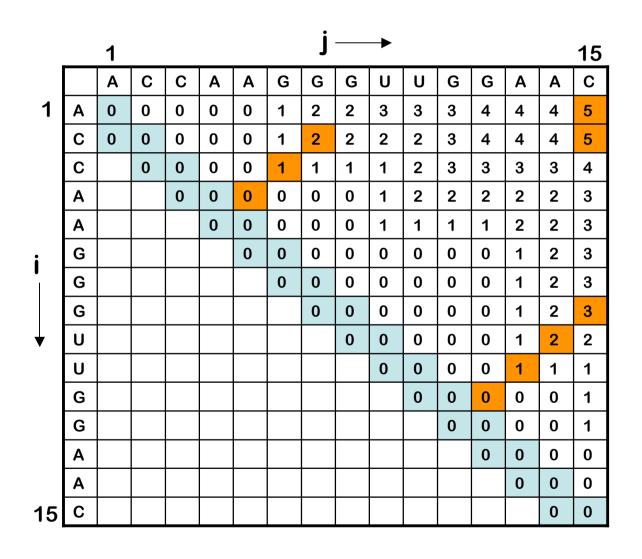
Only count G-C and A-U base pairs score = 1

How many base pairs does the optimal solution contain?

Which base pairs are present in the optimal solution?



Solution



Initialization

Traceback

Number of base pairs in optimal solution: 5

Actual base pairs: [(2,7),(3,6),(8,15), (9,14),(10,13)]

Thermodynamics

- $\Delta G = \Delta H T \Delta S$ ΔH is enthalpy, ΔS is entropy, and T is the temperature in Kelvin.
- Molecular interactions, such as hydrogen bonds, van der Waals and electrostatic interactions contribute to the ΔH term. ΔS describes the change of order of the system.
- Thus, both molecular interactions as well as the order of the system determine the direction of a chemical process.
- For any nucleic acid solution, it is extremely difficult to calculate the free energy from first principle
- Biophysical methods can be used to measure free energy changes

Thermodynamics

- Gibbs Free Energy, G
- Describes the energetics of biomolecules in aqueous solution. The change in free energy, ΔG , for a chemical process, such as nucleic acid folding, can be used to determine the direction of the process:
- $\Delta G=0$: equilibrium
- $\Delta G > 0$: unfavorable process
- ∆*G*<0: favorable process
- Thus the natural tendency for biomolecules in solution is to minimize free energy of the entire system (biomolecules + solvent).

RNA folding

Equilibrium between strands in folded/unfolded state. Lowest free energy structure is the most represented conformation at equilibrium

Free energy minimization (MFE)

Nearest-neighbor rules: free energies assigned to base pair stacks and to loops (unpaired regions). In helices, energy contributions depend on a base pair and its adjacent pair.

$$\Delta G_{37 \text{ total}}^{\circ} = 0.5 - 2.1 - 3.4 - 2.4 - 2.5 + 5.4$$

= -4.5 kcal/mol

Energy parameters from: Mathews et al. 1999

Zuker & Stiegler 1981

Dynamic programming solution (O(N³)) Two matrices: W & V

W(i,j) = the minimum free energy of all possible admissible structures formed from the subsequence S(i,j)
V(i,j) = the minimum free energy of all possible admissible structures formed from S(I,j) in which Si and Sj base pair with each other.

Implementations: mFold (now unafold) RNAfold (in Vienna RNA package)

Limitations

The accuracy of MFE methods is limited:

- -- free energy nearest-neighbor model is incomplete (e.g. motifs)
- -- some effects on stability are non-nearest-neighbor (bulge loops and single non-canonical pairs)
- -- not all RNA sequences are at equilibrium (i.e. kinetics might also be important!)
- -- Topological limitations (e.g. no pseudoknots!)
- -- RNA sequence might have multiple conformations (e.g. riboswitches, tRNA)

Need for suboptimal structure prediction:

RNAsubopt

- -- For sequence length N ~ 1.8N structures
- -- Sequence 100 nucleotides $\sim 3x10^{25}$ structures. Suppose calculate 1000 structs/sec it would take 10^{15} years!!!
- -- Zuker & Stiegler 1981, Mathews et al. 1999: heuristic set of suboptimal structures
- -- Wuchty 1999: all suboptimal structures within an energy increment above the MFE

Partition function

McCaskill 1990

A partition function is a quantity that encodes the statistical properties of a system in thermodynamic equilibrium. It connects the mechanics to thermodynamics

$$Q = \sum e^{-\Delta G/RT}$$

Sum over all possible secondary structures

R = gas constant

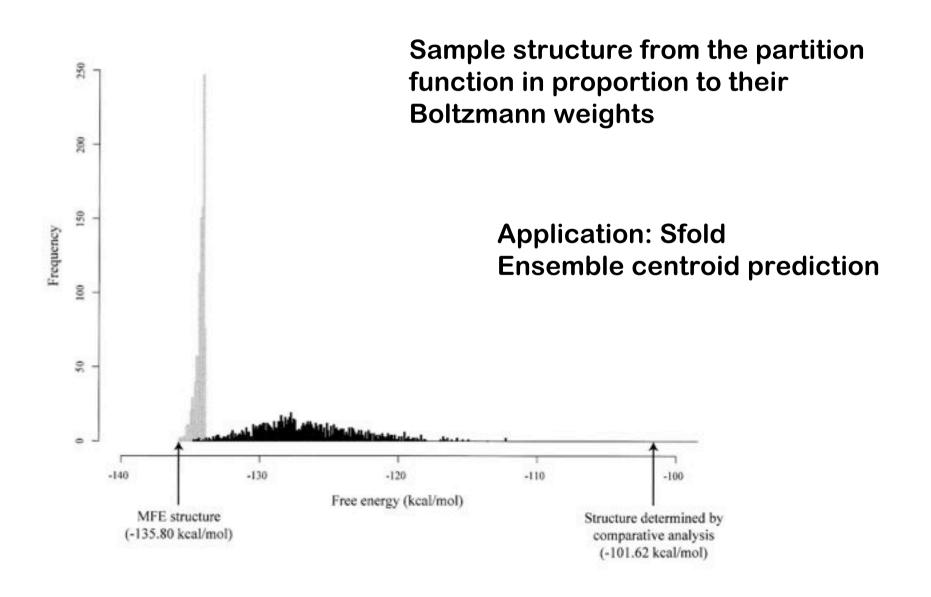
T = absolute temperature

Given the partition function, the probability of a given base pair is:

$$P = rac{\sum e^{-\Delta G/RT}}{Q}$$
 Sum over all secondary structures that contain the given base pair

Statistical sampling

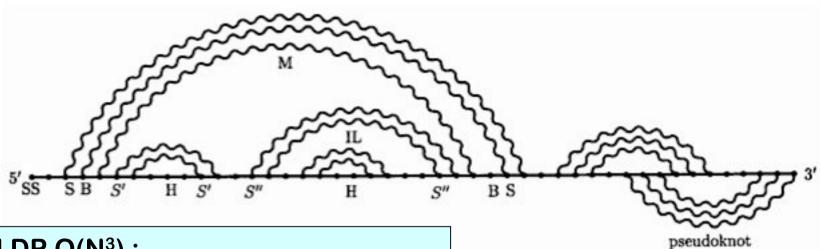
Ding & Lawrence 2003



Pseudoknot prediction

PKNOTS: E. Rivas & S.R. Eddy 1999

Two matrices vx and wx, like normal MFE Complex recursion rules to include pseudoknots Complexity: $O(N^6)$



Normal DP $O(N^3)$:

2x sequence length = 8x computer time Pseudoknots $O(N^6)$:

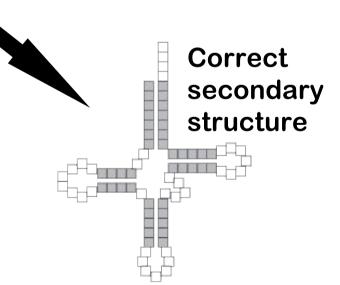
2x sequence length = 64x computer time!

Other approximations to predict pseudoknots

Comparative approaches

Input = multiple related sequences (homologues)
Two steps:
-- sequence alignment

-- prediction of common structure for all sequences



Covariation

Covariation, detectable at the sequence level, is caused by compensatory mutations (mostly isosteric base pairs)

Mutual information

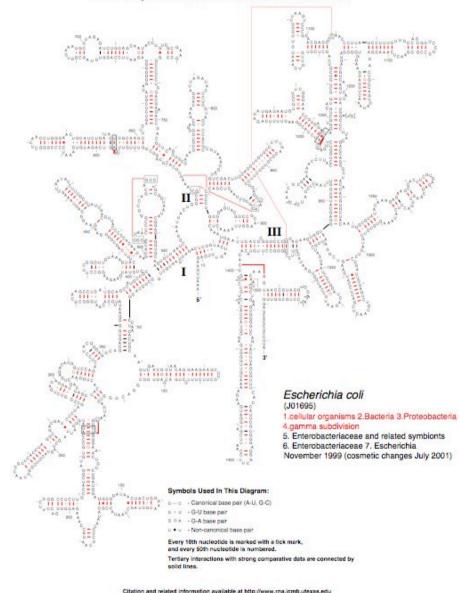
$$H(i,j) = F_{i,j}(N_1, N_2)log_2 \frac{F_{i,j}(N_1, N_2)}{f_i(N_1)f_j(N_2)}$$

$$N1, N2 \in U, C, A, G$$

$$F_{i,j}(N_1,N_2)$$
 Frequency of the (N1,N2) nucleotide pair in columns i and j

$$f_i(N_1), f_j(N_2)$$
 Frequencies of nucleotides N1 and N2 in columns i and j

Secondary Structure: small subunit ribosomal RNA



Ribosomal RNA

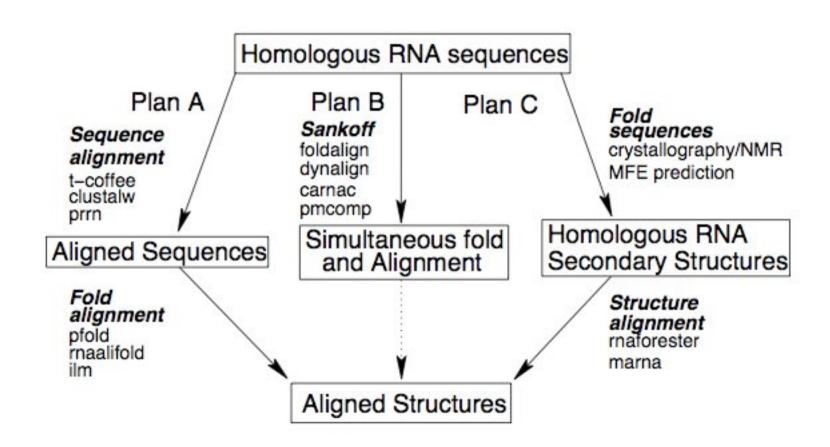
Large-scale alignment and covariation analysis led to accurate prediction of the SSU/LSU rRNA structures.

Using the recently solved crystal structures of the ribosome, the models were shown to be over 97% accurate (Gutell et al. 2002)

Comparative RNA web Cannone et al. 2002

Several approaches:

- 1) Simultaneously align and fold
- 2) Predict secondary structure for a given alignment
- 3) Fold sequences, align structures



Align and fold

D. Sankoff 1985

Simultaneous RNA sequence alignment and folding Theoretically ideal, but computationally over-expensive Complexity = $O(n^{3m})$ time and $O(n^{2m})$ in space n = sequence leghth and m = number of sequences

Many approximations exist making the Sankoff method practical (e.g. FOLDALIGN and Dynalign)

FOLDALIGN: Gorodkin et al. 1997
Limits the maximum distance between paired nucleotides and the maximum length difference for fragments being aligned

Dynalign: Mathews & Turner 2002
Limits the maximum difference in index for a nucleotide in the first sequence that will be aligned to the second sequence

Predict from alignment

RNAalifold: Hofacker et al. 2002 Extension to Zuker-Stiegler algorithm for computing a consensus structure from RNA alighments Combines thermodynamics and covariation

> Pfold: Knudsen & Hein 1999, 2003 Stochastic context free grammar (SCFG)

```
S \rightarrow aSu \mid uSa \mid cSg \mid gSc

S \rightarrow aS \mid cS \mid gS \mid uS

S \rightarrow Sa \mid Sc \mid Sg \mid Su

S \rightarrow SS R. Dowell & S.R. Eddy 2004

S \rightarrow \varepsilon
```

Evaluation of several lightweight stochastic context-free grammars for RNA secondary structure prediction

Fold and align structures

RNAforester: Hochsmann et al 2003

Tree alignment model

Pairwise alignment of two input structures

Consensus shapes: J. Reeder & R. Giegerich 2005 Abstract shapes: representation of RNA secondary structure that displays the branching pattern of the helices

```
(((...((...(((...)))))).((...))))
    Level 5 shape: [[][]]
    Level 3 shape: [[[]][]]
```

- 1) Generate list of abstract shapes (much shorter than the list of suboptimal structures within a certain energy range)
- 2) Find the lowest free energy abstract shape common to all sequences

Combine multiple sources

RNAstructure: Mathews et al. 2004 Combines thermodynamics, covariation, chemical modification constraints

Welcome to BayesFold

Get Started:

- Fold an alignment
- Read the help
- · Download the manual

Contact Us:

- Suggestions
- · Bug reports
- Support requests

Learn More:

BayesFold poster

The current version is BayesFold 1.01

Note about jaynes: You should only access BayesFold through the server at bayes.colorado.edu. If you see jaynes.colorado.edu in the address bar, you are using an internal development site that is often under construction. Click **here** for the correct link. BayesFold will typically not work on the jaynes.colorado.edu server!

NEWS: Based on requests from users, two new features have been added to BayesFold. Structures may now be drawn in either the clockwise (default) or counterclockwise direction, and unpaired structure ends are drawn spread slightly apart to make ends easier to identify. Details of these new features are available in the manual.

BayesFold is a web application that finds, ranks, and draws the likeliest structures for a sequence alignment. Foldings are based on the predictions of the Bayesian statistical method. BayesFold provides convenient structure comparison and formatting functionality, and produces publication-quality graphics.

To use BayesFold, your browser must be Internet Explorer 5+ for Windows with the Adobe

s Viewer plugin installed. When you access BayesFold, the program browser configuration and tell you about any problems. Future browsers as browser XML and SVG standards support improves.

nd maintained by Rob Knight and Amanda Birmingham at the Boulder. Should you encounter any difficulties or errors while using ge you to email us about the problem.

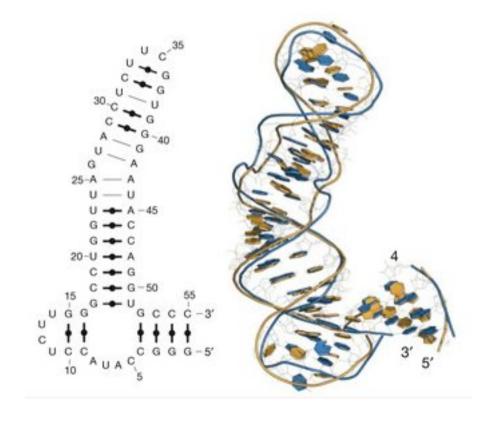
BayesFold: Knight et al. 2004 Combines thermodynamics, covariation, experimental constraints

Latest greatest

M. Parisien and F. Major. Nature (2008). 452:51-55. The MC-Fold and MC-Sym pipeline infers RNA structure from sequence data

Structure prediction pipeline

- -- secondary structure
- -- tertiary structure
 Based on nucleotide cyclic
 motifs (NCMs)
 Includes all base pairs (not only
 canonical pairs)



Accuracy

Benchmark study: Garder & Giegerich 2004

Accuracy metrics:

True/False positives, True/False negatives
Sensitivity, positive predictive value, Matthews
correlation coefficient, and others

Comparative approaches in general more accurate than single-sequence MFE methods.

- -- Several factors limit the accuracy of MFE methods (see earlier slide).
- -- Structure & function are more conserved than sequence.
 Multiple sequences provide many constraints on the structure.

Summary

Experimental approaches

- -- several experimental techniques are available, ranging in precision
- -- experiments are difficult, expensive, and time-consuming
- -- if they work, they are a great source of structural information

Computational methods are necessary

- -- single-sequence versus multi-sequence methods
- -- minimum free energy prediction methods
- -- suboptimal foldings, partition function, and statistical sampling
- -- comparative approaches, three strategies
- -- integration of multiple sources of evidence