#### Helices of RNAs

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#### Outline

- Introduction
- Development of a new structure abstraction
- Implementation of an algorithm based on the new abstraction
- Possible problems
- Evaluation of the algorithm
- Designing RNA class predictors

## Structural components of RNA

#### RNA has different structural components:

- single-stranded regions (SS)
- hairpin loops (HL)
- stacking regions (SR)
- bulges on 5´side (BL) or 3´side (BR)
- internal loops (IL)
- multiloops (ML).

## Structural components of RNA

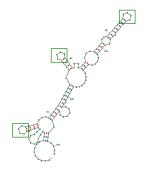
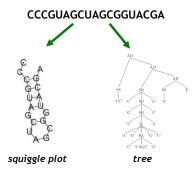


Figure: Structural components



## **Suboptimal structures**

- But the 84true" structure is not always the one with
- the lowest predicted free energy.

## **Introducing abstract shapes**

• Solution: Use abstract shapes to describe a set of structures.

## **Introducing abstract shapes**

- In the domain of shapes, we care only about
- open structures (OP)
- closed structure (CP)
- branching ("fork", FK) and
- adjacency of structures (AD).

## **Defining abstract shapes**

- A (very abstract) abstraction function :
- $\pi(SS(I)) = OP$  (single-stacked region)
- $\pi(HL(a,l,b) = CL \text{ (hairpin)}$
- $\pi(SR(a,x,b)) = \pi(x)$  (stacked region)
- $\pi(IL(a,l,x,lb4,b)) = f0(x)$  (interior loop)
- $\pi(ML(a,c,b)) = FK(\pi(x))$  (multiloop)
- E represents the "empty structure".
- where a, b = nucleotides, l = loop, c = list of adjacent
- components and x = arbitrary structure elements.

# Example

- Abstract shapes are a homomorphic image of the folding
- space of a RNA sequence (same as mfe or base pair
- maximization).
- They can therefore be computed using Dynamic
- Programming (DP).
- Giegerich et al. use an extension of DP called
- Algebraic Dynamic Programming (ADP).
- ADP offers some interesting aspects (separation of
- recognition and evaluation, pair algebras).

## **Computing abstract shapes**

• Algebraic Dynamic Programming defines the set of all possible solutions (e.g. foldings) using a context-free grammar.

### **First summary**

- One abstract shape represents a family of
- similar RNA structures.
- Shapes are defined by an abstraction function
- that maps from structure to shape space.
- Shapes can be used to represent a large number of (suboptimal) foldings to obtain an holistic view of the folding space.

- Suboptimal Folding
- Out of 99 tRNA sequences in Rfam only 30 had the typical cloverleaf structure as predicted mfe folding.
- Example: tRNA of Natronobacterium pharaonis:
- mfe structure is a hairpin with internal loops, cloverleaf structure occurs at position 104 of 199
- suboptimal structures.
- All these suboptimal folding can be represented by three abstract shapes.

- Suboptimal Folding
- (demo)

# **Suboptimal folding**

- Better structure prediction (than mfe folding) can
- be obtained using comparative approaches:

• A possible resort :

• Consensus structures with abstract shapes

• Possible scoring functions:

## Going comparative pays off

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# **Comparison with Sankoff**

## Summary

- Abstract shapes represent disjoint classes of RNA foldings.
- Shapes are computed using (Algebraic) Dynamic Programming.
- Inspecting the abstract shape space of a sequence
- can give a quick overview of the folding space.
- Consensus folding with abstract shapes performs well.
- Choice of best abstraction function and energy range
- is important but difficult.

#### **Conclusions and Outlook**

- Other approaches to suboptimal folding exist such as statistical sampling of the folding space.
- Text representations of shapes could be used as index in structure databases to classify non-coding RNA.
- Extensions to the shape formalisms are under work
- ( e.g. computation of shape probabilities, de novo prediction of non-coding RNA genes )

### End

- Thanks a lot for your attention !
- Questions ?