

M2 BIM/STRUCT - Lecture 1

Folding RNA *in silico*

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1 Introduction

- Dynamic programming 101
- Why RNA?
- RNA folding
- RNA Structure(s)
- Some representations of RNA structure

2 Some flavours of folding prediction

- Thermodynamics vs Kinetics
- Dynamic programming: Reminder

3 Free-energy minimization

- Nussinov-style RNA folding
- Turner energy model
- MFold/Unafold
- Performances and the comparative approach
- Towards a 3D ab-initio prediction

... or how to make a million bucks by giving change parsimoniously!!

Problem: You have access to unlimited amount of **1**, **20** and **50** cents coins. A client prefers to travel light, i.e. to **minimize the #coins**. How to give **N** cents back in change without losing a customer?

Strategy #1: Start with *heaviest* coins, and then complete/fill-up with coins of *decreasing* value.

21 = ??

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60

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$$= \text{€20} + \text{€20} + \text{€20} !$$

Problem *a priori* (!) non-solvable using such a *greedy* approach, as a (simpler) problem is already NP-complete (thus Efficient solution \Rightarrow 1M\$).

Strategy #2: Brute force enumeration $\rightarrow \#Coins^N$ (Ouch!)

Strategy #3: The following recurrence gives the minimal number of coins:

$$Min\#Coins(N) = \text{Min} \left\{ \begin{array}{l} \text{1€} \rightarrow 1 + Min\#Coins(N - 1) \\ \text{2€} \rightarrow 1 + Min\#Coins(N - 20) \\ \text{5€} \rightarrow 1 + Min\#Coins(N - 50) \end{array} \right.$$

With some memory (N intermediate computations), the minimum number of coins can be obtained after $N \times \#Coins$ operations. An actual set of coins can be reconstructing by **tracing back** the choices performed at each stage, leading to the minimum.

Remark: We still haven't won the million, as N has **exponential value compared to the length of its encoding**, so the algorithm does not qualify as *efficient* (i.e. polynomial).

Still, this approach is much more efficient than a brute-force enumeration:
 \Rightarrow Dynamic programming.

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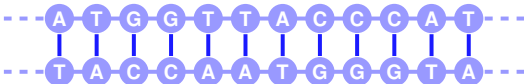
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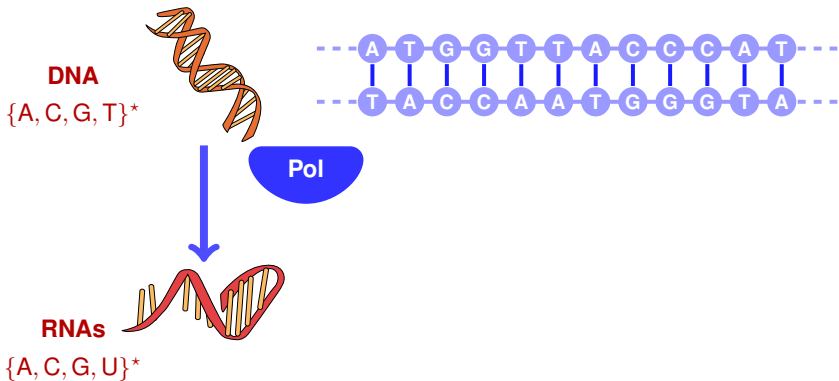
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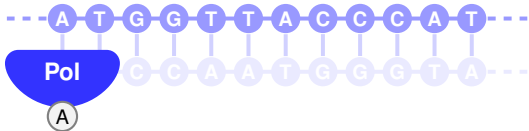
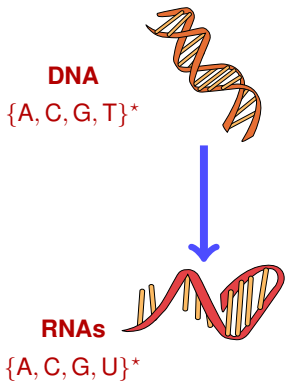
DNA
{A, C, G, T}^{*}



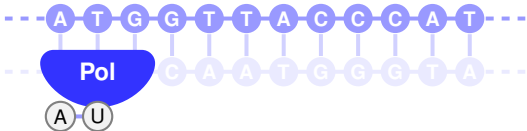
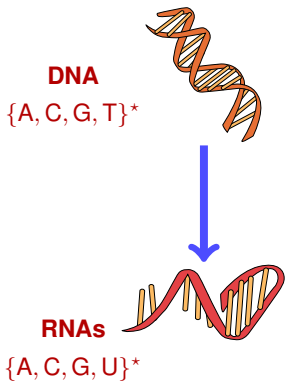
Fundamental *dogma* of molecular biology



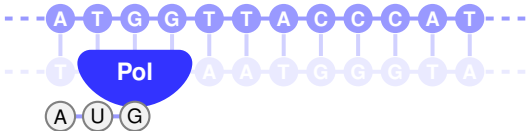
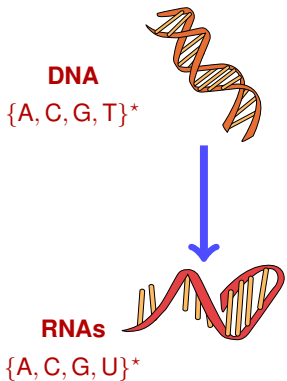
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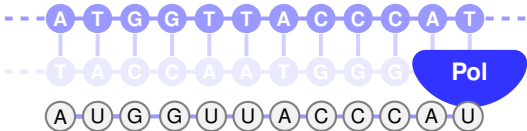
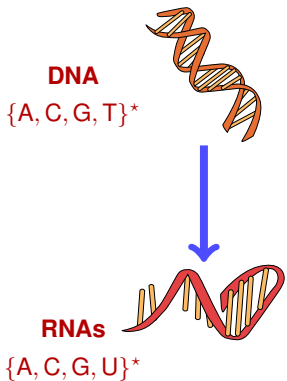
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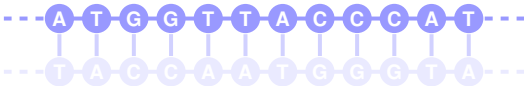
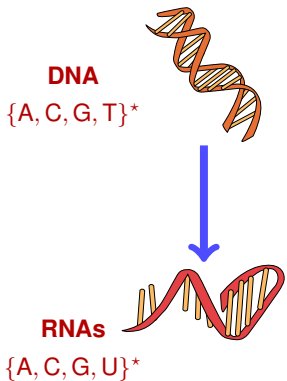
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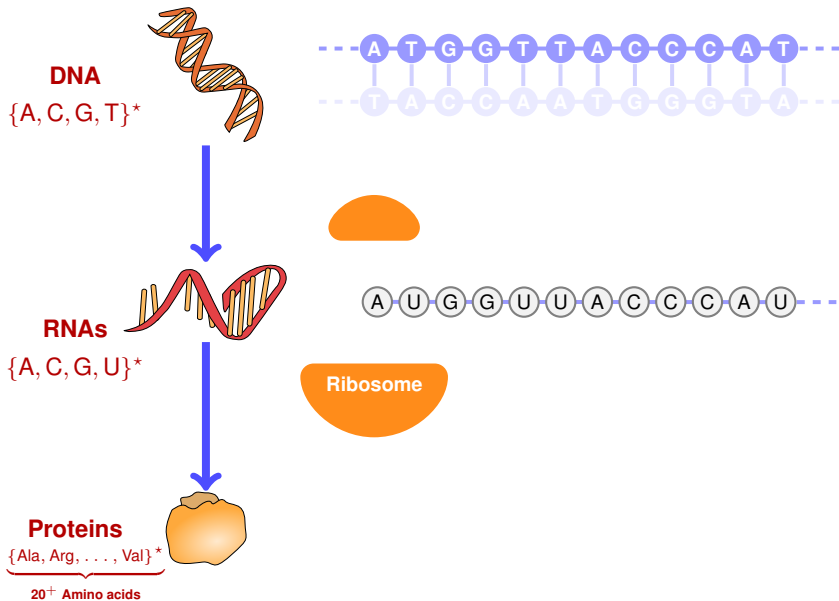
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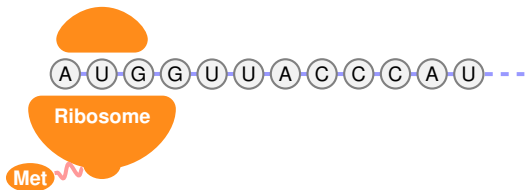
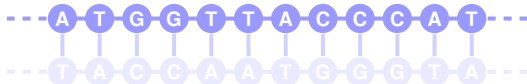
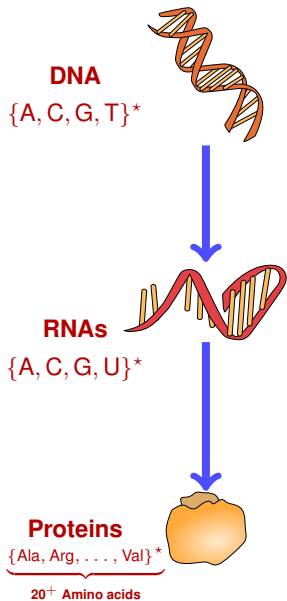
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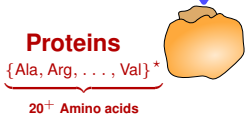
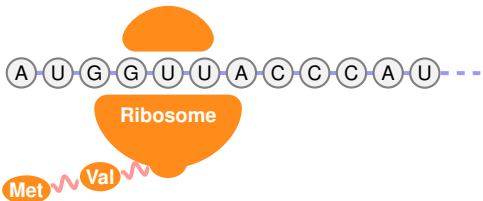
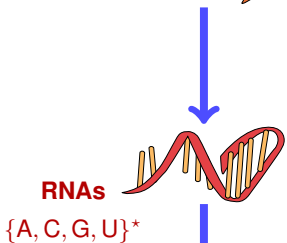
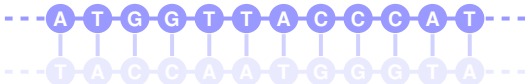
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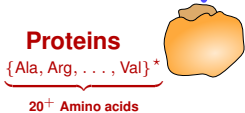
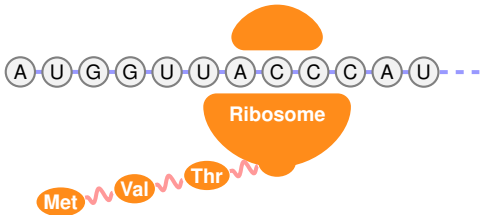
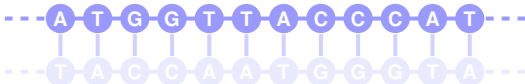
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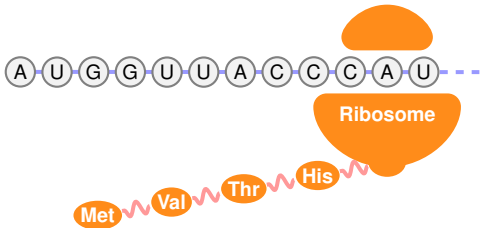
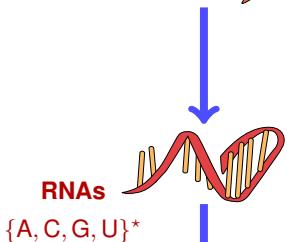
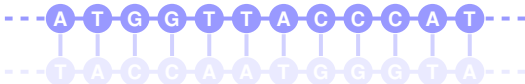
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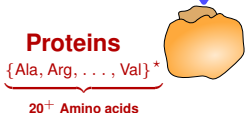
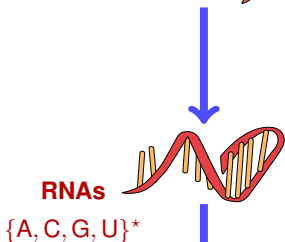
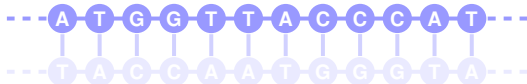
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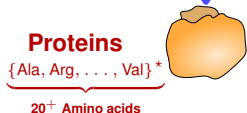
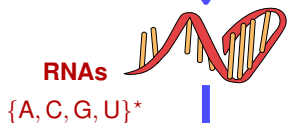
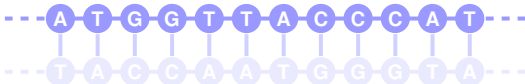
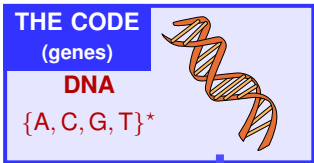
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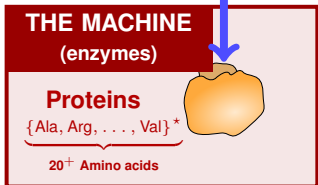
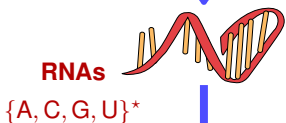
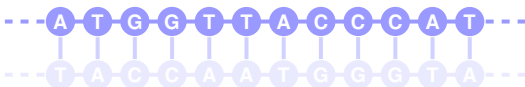
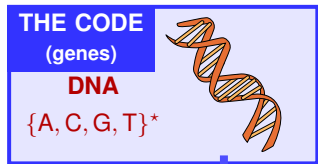
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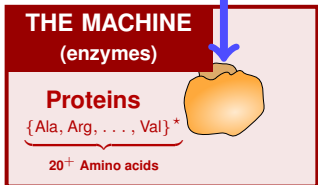
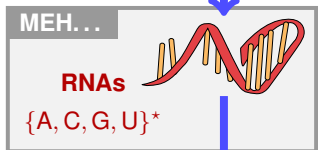
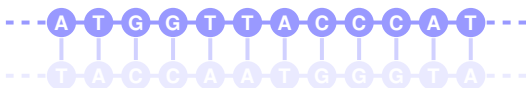
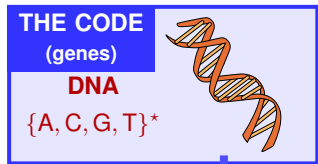
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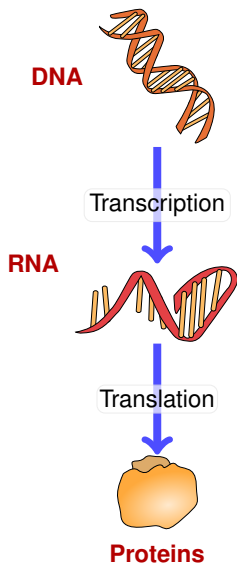


Fundamental dogma of molecular biology

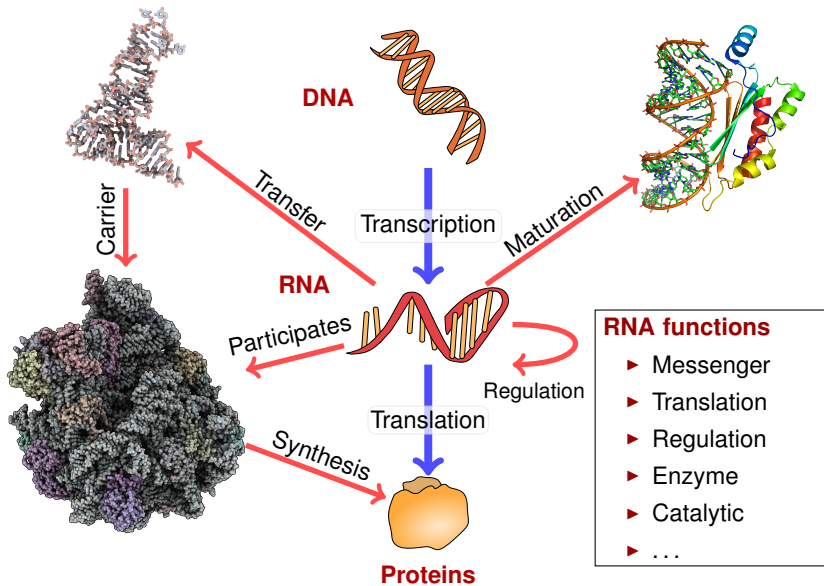


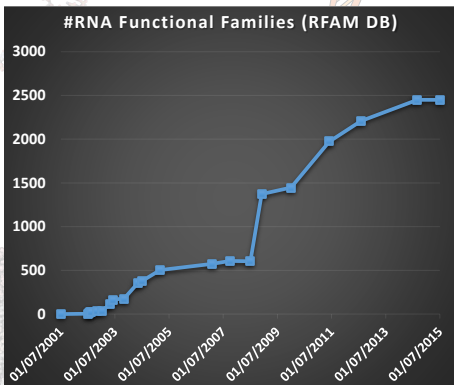
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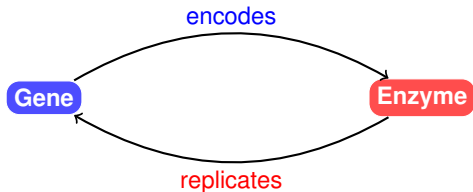




RNA functions

- ▶ Messenger
- ▶ Translation
- ▶ Regulation
- ▶ Enzyme
- ▶ Catalytic
- ▶ ...

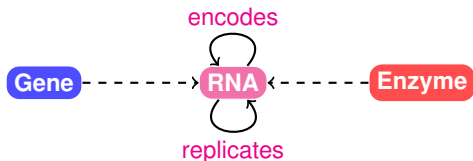
Proteins



A **gene** big enough to specify **an enzyme** would be too big to replicate accurately without the aid of **an enzyme** of the very kind that it is trying to specify. So the system *apparently cannot get started*.

[...] This is the **RNA World**. To see how plausible it is, we need to look at why proteins are good at being enzymes but bad at being replicators; at why DNA is good at replicating but bad at being an enzyme; and finally why **RNA might just be good enough at both roles to break out of the Catch-22**.

R. Dawkins. *The Ancestor's Tale: A Pilgrimage to the Dawn of Evolution*



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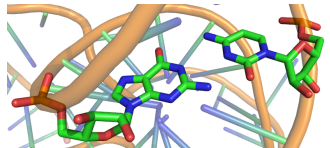
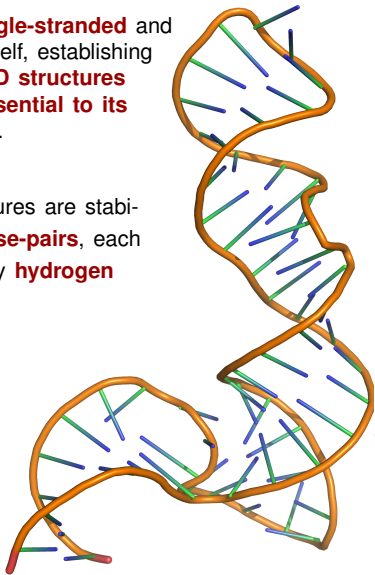
[...] This is the **RNA World**. To see how plausible it is, we need to look at why proteins are good at being enzymes but bad at being replicators; at why DNA is good at replicating but bad at being an enzyme; and finally why **RNA might just be good enough at both roles to break out of the Catch-22**.

R. Dawkins. *The Ancestor's Tale: A Pilgrimage to the Dawn of Evolution*

RNA folding

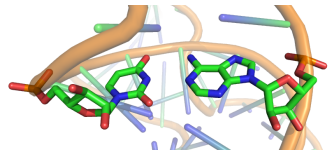
RNA is **single-stranded** and **folds** on itself, establishing **complex 3D structures** that are **essential to its function(s)**.

RNA structures are stabilized by **base-pairs**, each mediated by **hydrogen bonds**.

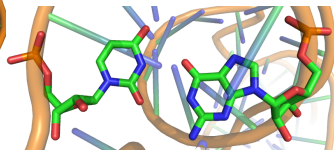


G/C

Watson/Crick base-pairs



U/A



U/G

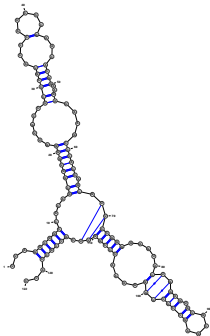
Wobble base-pair

Canonical base-pairs

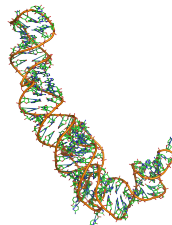
Three¹ levels of representation:

```
UUAGGCGGCCACAGC  
GGUGGGGUUGCCUCC  
CGUACCCAUCCGAA  
CACGGAAGAUAGCC  
CACCAGCGUCCGGG  
GAGUACUGGAGUGCG  
CGAGCCUCUGGGAAA  
CCCGGUUCGCCCA  
CC
```

Primary structure



Secondary structure



Tertiary structure

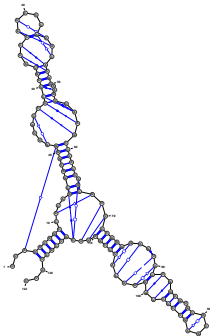
Source: 5s rRNA (PDB 1K73:B)

¹Well, mostly...

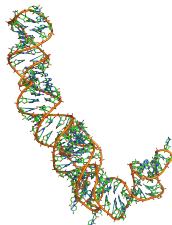
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CACGGAAGAUAGCC  
CACCAGCGUCCGGG  
GAGUACUGGAGUGCG  
CGAGCCUCUGGGAAA  
CCCGGUUCGCCGCA  
CC
```

Primary structure



Secondary⁺ structure



Tertiary structure

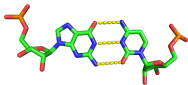
Source: 5s rRNA (PDB 1K73:B)

¹Well, mostly...

► Non-canonical base-pairs

Any base-pair **other than** {(A-U), (C-G), (G-U)}

Or interacting on non-standard edge (\neq WC/WC-Cis) [?].

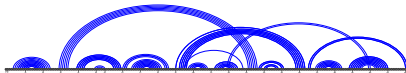


Canonique CG pair(WC/WC-Cis)



Non-canonique CG pair (Sugar/WC-Trans)

► Pseudoknots (PKs)



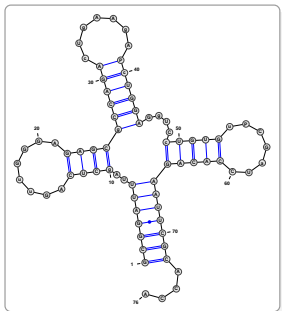
Pseudoknotted structure of group I ribozyme (PDBID: 1Y0Q:A)

Considering PKs may lead to better predictions, **but:**

- Some PK conformations are simply unfeasible;
- Folding *in silico* with general pseudoknots is NP-complete [?];

Still, folding on restricted classes of conformations seems promising [?].

Various representations for a versatile biomolecule



Outer-planar graphs

Hamiltonian-path, $\Delta(G) \leq 3$, 2-connected*

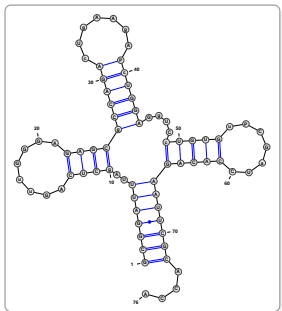
Supporting intuitions

Different representations

Common combinatorial structure

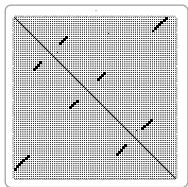
* Additional steric constraints

Various representations for a versatile biomolecule



Outer-planar graphs

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Dot plots

Adjacency matrices*

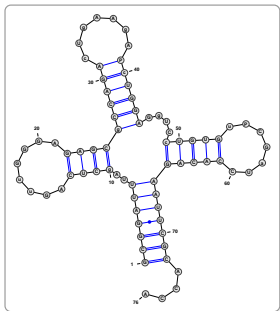
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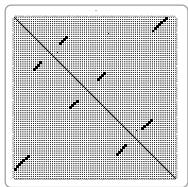
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Various representations for a versatile biomolecule

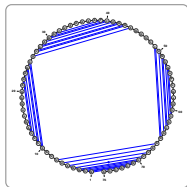


Outer-planar graphs

Hamiltonian-path, $\Delta(G) \leq 3$, 2-connected*



Dot plots
Adjacency matrices*



Non-crossing arc diagrams*

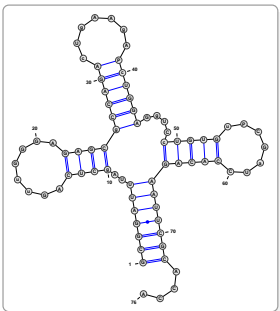
Supporting intuitions

Different representations

Common combinatorial structure

* Additional steric constraints

Various representations for a versatile biomolecule

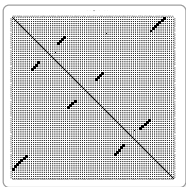


Outer-planar graphs

Hamiltonian-path, $\Delta(G) \leq 3$, 2-connected*

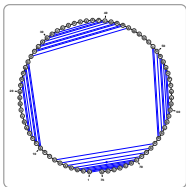
(((((((.....)))))).....)))).....

Motzkin words*



Dot plots

Adjacency matrices*



Non-crossing arc diagrams*

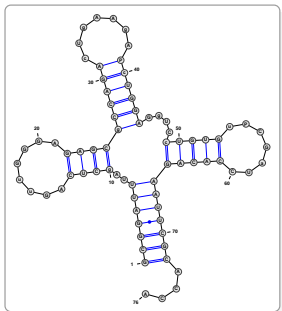
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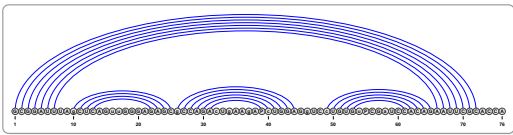


Outer-planar graphs

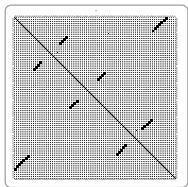
Hamiltonian-path, $\Delta(G) \leq 3$, 2-connected*

(((((((.....))))))(((((((.....)))))).....(((((((.....))))))))).....

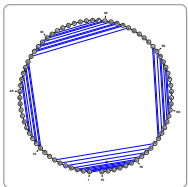
Motzkin words*



Non-crossing arc-annotated sequences*



Dot plots
Adjacency matrices*



Non-crossing arc diagrams*

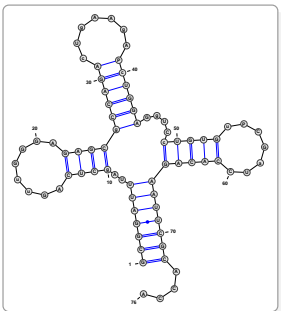
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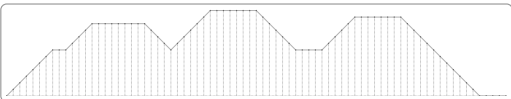


Outer-planar graphs

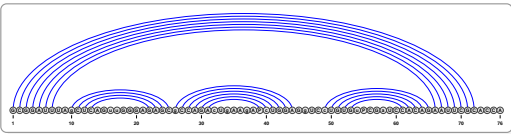
Hamiltonian-path, $\Delta(G) \leq 3$, 2-connected*

(((((((.....))))))(((.....)))).....((((.....)))))))).....

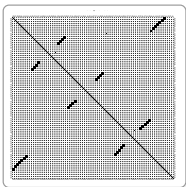
Motzkin words*



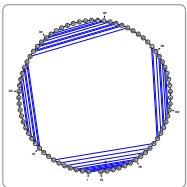
Positive 1D meanders* over $S = \{+1, -1, 0\}$



Non-crossing arc-annotated sequences*



Dot plots
Adjacency matrices*



Non-crossing arc diagrams*

Supporting intuitions

Different representations

Common combinatorial structure

* Additional steric constraints

1 Introduction

- Dynamic programming 101
- Why RNA?
- RNA folding
- RNA Structure(s)
- Some representations of RNA structure

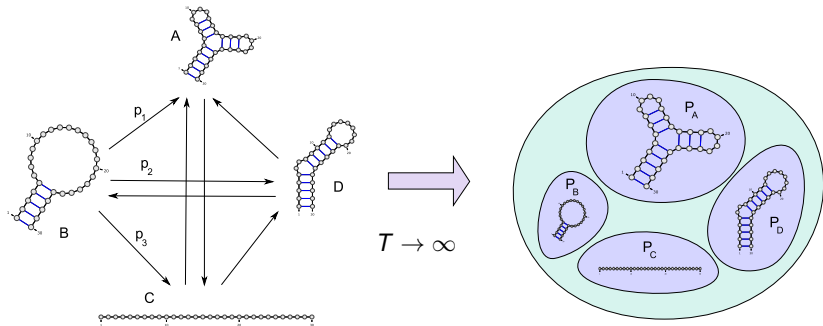
2 Some flavours of folding prediction

- Thermodynamics vs Kinetics
- Dynamic programming: Reminder

3 Free-energy minimization

- Nussinov-style RNA folding
- Turner energy model
- MFold/Unafold
- Performances and the comparative approach
- Towards a 3D ab-initio prediction

At the nanoscopic scale, RNA structure *fluctuates* (\approx Markov process).



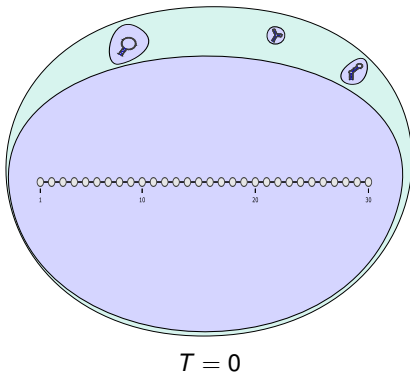
Convergence towards a **stationary distribution** at the **Boltzmann equilibrium**, where the probability of a conformation only depends on its **free-energy**.

Corollary: Initial conformation does not matter.

Questions: For a given **conformation space** and **free-energy** model:

- Determine most stable (Minimum Free-Energy) structure at equilibrium;
- Compute average properties of Boltzmann ensemble;

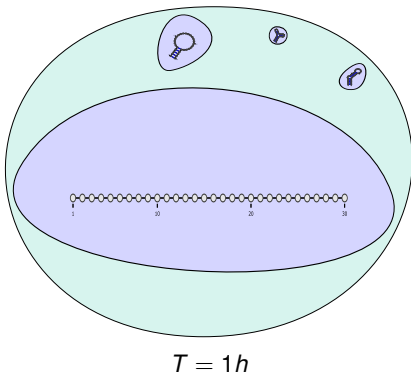
Transcription: RNA synthesized, supposedly without structure²



But most mRNAs are degraded before 7h (Org.: Souris [?]).

²Except for co-transcriptional folding...

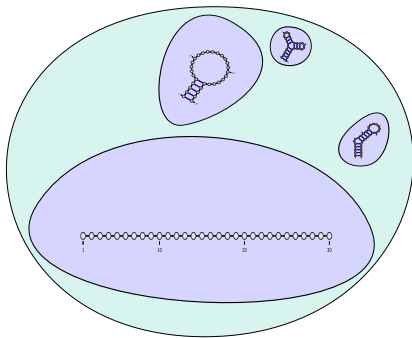
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But most mRNAs are degraded before 7h (Org.: Souris [?]).

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Transcription: RNA synthesized, supposedly without structure²

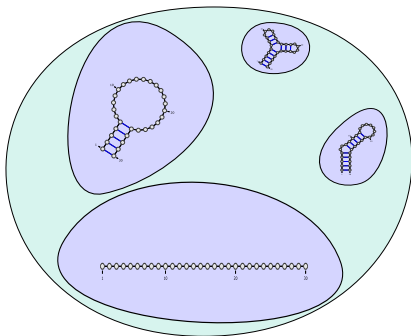


$T = 2h$

But most mRNAs are degraded before 7h (Org.: Souris [?]).

²Except for co-transcriptional folding...

Transcription: RNA synthesized, supposedly without structure²

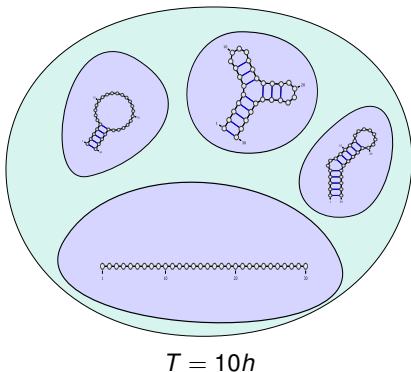


$T = 5h$

But most mRNAs are degraded before 7h (Org.: Souris [?]).

²Except for co-transcriptional folding...

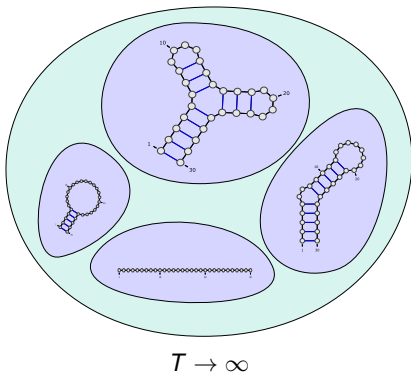
Transcription: RNA synthesized, supposedly without structure²



But most mRNAs are degraded before 7h (Org.: Souris [?]).

²Except for co-transcriptional folding...

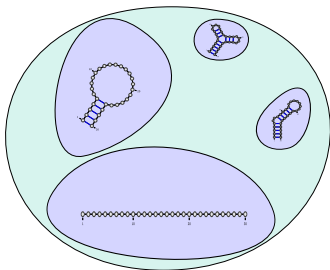
Transcription: RNA synthesized, supposedly without structure²



But most mRNAs are degraded before 7h (Org.: Souris [?]).

²Except for co-transcriptional folding...

Transcription: RNA synthesized, supposedly without structure²



$T = 10h$

But most mRNAs are degraded before 7h (Org.: Souris [?]).

- A.** Determine most stable (Minimum Free-Energy) structure at equilibrium;
- B.** Compute average properties of Boltzmann ensemble;
- C. Determine most likely structure at finite time T .**
(c.f. H. Isambert through simulation, NP-complete deterministically [?])

²Except for co-transcriptional folding...

Dynamic programming = General optimization technique.

Prerequisite: Optimal solution for problem P can be derived from solutions to strict sub-problems of P .

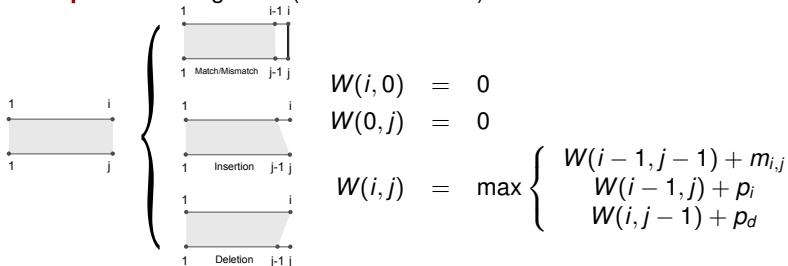
Bioinformatics :

Discete solution space (alignments, structures...)

+ Additively-inherited objective function (cost, log-odd score, energy...)

⇒ Efficient dynamic programming scheme

Example: Local Alignment(Smith/Waterman)



Dynamic programming scheme defines a space of (sub)problems and a recurrence that relates the score of a problem to that of smaller problems.

Given a scheme, two steps :

- ▶ **Matrix filling**: Computation and tabulation of best scores (Computed from smaller problems to larger ones).
- ▶ **Traceback**: Reconstruct best solution from contributing subproblems.

Complexity of algorithm depends on:

- ▶ **Cardinality** of sub-problem space
- ▶ **Number of alternatives** considers at each step (#Terms in recurrence)

Smith&Waterman example:

- ▶ $i: 1 \rightarrow n + 1 \Rightarrow \Theta(n)$
- ▶ $j: 1 \rightarrow m + 1 \Rightarrow \Theta(m)$
- ▶ 3 operations at each step

$\Rightarrow \Theta(m.n)$ time/memory

$$W(i, 0) = 0$$

$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

Complete example

Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

$$W(i, 0) = 0$$

$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

		A	C	A	C	A	C	T	A
	0	0	0	0	0	0	0	0	0
A	0								
G	0								
C	0								
A	0								
C	0								
A	0								
C	0								
A	0								

Complete example

Example: Local alignment of AGCACACA and ACACACTA

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	A	C	A	C	A	C	T	A
	0	0	0	0	0	0	0	0
A	0	2						
G	0							
C	0							
A	0							
C	0							
A	0							
C	0							
A	0							

Complete example

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Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

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		A	C	A	C	A	C	T	A
	0	0	0	0	0	0	0	0	0
A	0	2	1						
G	0								
C	0								
A	0								
C	0								
A	0								
C	0								
A	0								

Diagram illustrating the dynamic programming table for local alignment. The table shows the alignment of AGCACACA (rows) and ACACACTA (columns). The value 2 is highlighted in the cell (A, C), and arrows indicate the path from (0,0) to (1,2) and then to (1,3).

Complete example

Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

$$W(i, 0) = 0$$

$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

		A	C	A	C	A	C	T	A
	0	0	0	0	0	0	0	0	0
A	0	2	1	2					
G	0								
C	0								
A	0								
C	0								
A	0								
C	0								
A	0								

Red arrows indicate the path: (0,0) to (1,1) to (1,2) to (2,3).

Complete example

Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

$$W(i, 0) = 0$$

$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

		A	C	A	C	A	C	T	A
	0	0	0	0	0	0	0	0	0
A	0	2	1	2	1				
G	0								
C	0								
A	0								
C	0								
A	0								
C	0								
A	0								

Complete example

Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

$$W(i, 0) = 0$$

$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

		A	C	A	C	A	C	T	A
	0	0	0	0	0	0	0	0	0
A	0	2	1	2	1	2	1	0	2
G	0								
C	0								
A	0								
C	0								
A	0								
C	0								
A	0								

Red arrows indicate the path of the local alignment: (0,0) to (1,1), (1,1) to (1,2), (1,2) to (2,3), (2,3) to (2,4), (2,4) to (3,5), (3,5) to (3,6), (3,6) to (3,7), (3,7) to (3,8), (3,8) to (3,9).

Complete example

Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

$$W(i, 0) = 0$$

$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

	A	C	A	C	A	C	T	A	
0	0	0	0	0	0	0	0	0	
A	0	2	1	2	1	2	1	0	2
G	0	1	1	1	1	1	1	0	1
C	0								
A	0								
C	0								
A	0								
C	0								
A	0								

Complete example

Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

$$W(i, 0) = 0$$

$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

	A	C	A	C	A	C	T	A	
0	0	0	0	0	0	0	0	0	
A	0	2	1	2	1	2	1	0	2
G	0	1	1	1	1	1	1	0	1
C	0	0	3	2	3	2	3	2	1
A	0								
C	0								
A	0								
C	0								
A	0								

Complete example

Example: Local alignment of AGCACACA and ACACA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

$$W(i, 0) = 0$$

$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

	A	C	A	C	A	C	T	A	
0	0	0	0	0	0	0	0	0	
A	0	2	1	2	1	2	1	0	2
G	0	1	1	1	1	1	1	0	1
C	0	0	3	2	3	2	3	2	1
A	0	2	2	5	4	5	4	3	4
C	0								
A	0								
C	0								
A	0								

Complete example

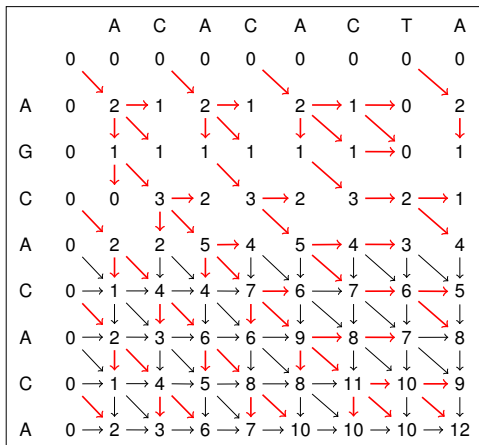
Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

$$W(i, 0) = 0$$

$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$



Complete example

Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

$$W(i, 0) = 0$$

$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

		A	C	A	C	A	C	T	A
0	0	0	0	0	0	0	0	0	0
A	0	2	1	2	1	2	1	0	2
G	0	1	1	1	1	1	1	0	1
C	0	0	3	2	3	2	3	2	1
A	0	2	2	5	4	5	4	3	4
C	0	1	4	4	7	6	7	6	5
A	0	2	3	6	6	9	8	7	8
C	0	1	4	5	8	8	11	10	9
A	0	2	3	6	7	10	10	10	12

Complete example

Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

$$W(i, 0) = 0$$

$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

Best alignment

		A	C	A	C	A	C	T	A
0	0	0	0	0	0	0	0	0	0
A	0	2	1	2	1	2	1	0	2
G	0	1	1	1	1	1	1	0	1
C	0	0	3	2	3	2	3	2	1
A	0	2	2	5	4	5	4	3	4
C	0	1	4	4	7	6	7	6	5
A	0	2	3	6	6	9	8	7	8
C	0	1	4	5	8	8	11	10	9
A	0	2	3	6	7	10	10	10	12

Complete example

Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

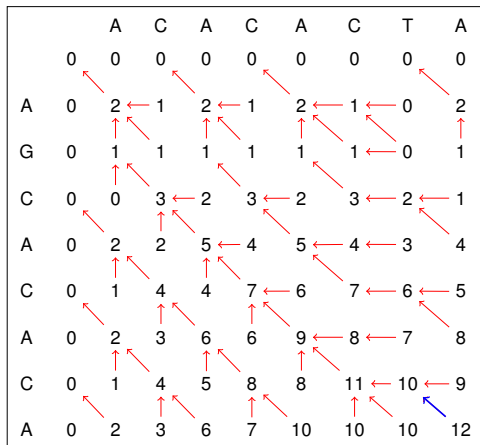
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$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

Best alignment

A
A



Complete example

Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

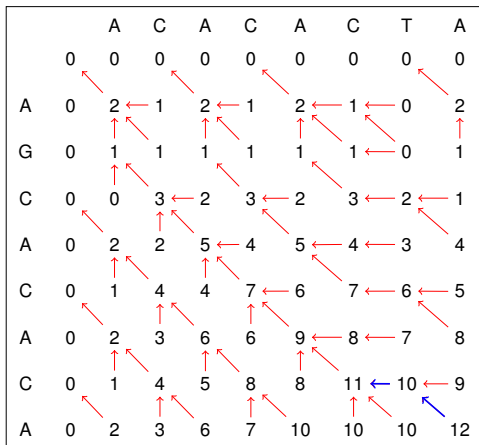
$$W(i, 0) = 0$$

$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

Best alignment

- A
T A



Complete example

Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

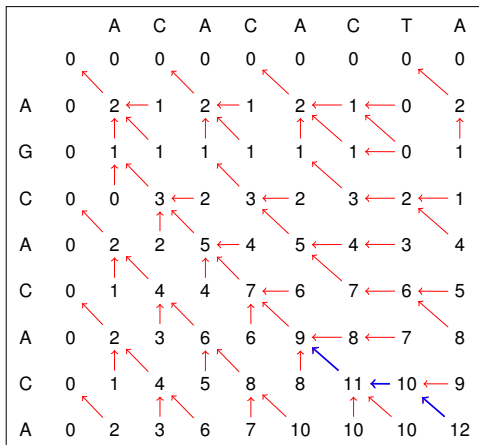
$$W(i, 0) = 0$$

$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

Best alignment

C - A
C T A



Complete example

Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

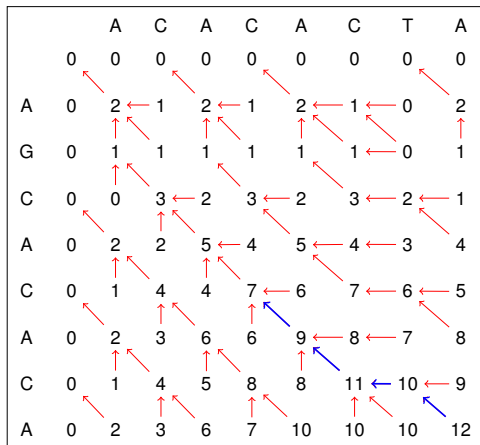
$$W(i, 0) = 0$$

$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

Best alignment

A C - A
A C T A



Complete example

Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

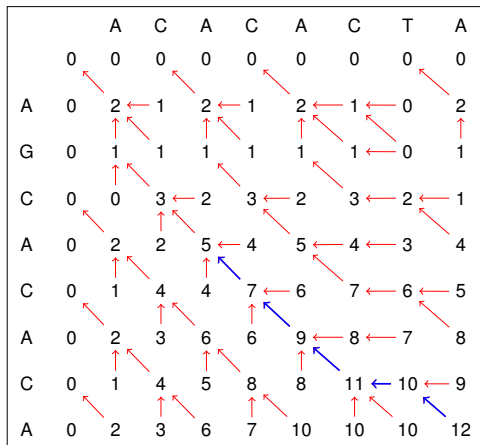
$$W(i, 0) = 0$$

$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

Best alignment

C A C - A
C A C T A



Complete example

Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

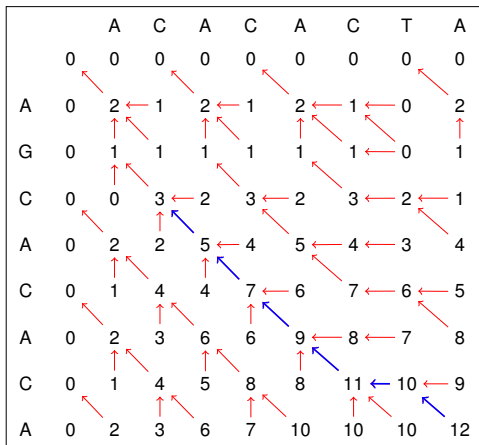
$$W(i, 0) = 0$$

$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

Best alignment

A	C	A	C	-	A
A	C	A	C	T	A



Complete example

Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

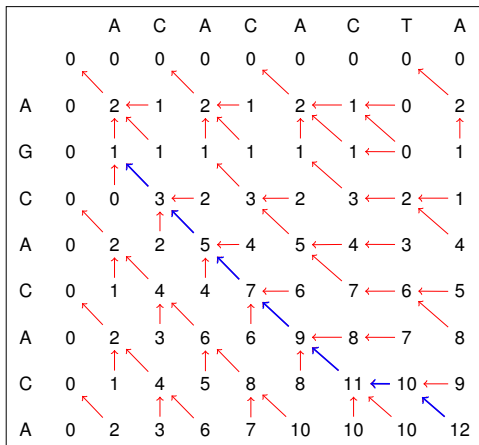
$$W(i, 0) = 0$$

$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

Best alignment

C	A	C	A	C	-	A
C	A	C	A	C	T	A



Complete example

Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

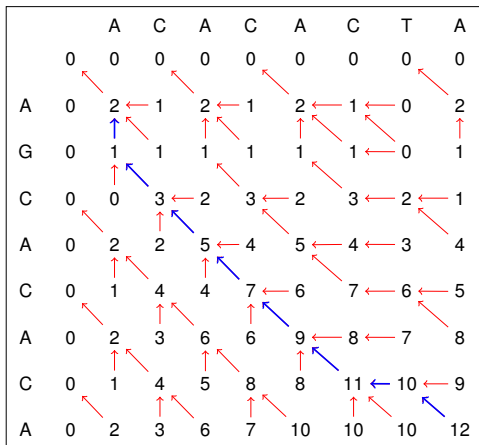
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$$W(0, j) = 0$$

$$W(i, j) = \max \begin{cases} W(i-1, j-1) + m_{i,j} \\ W(i-1, j) + p_i \\ W(i, j-1) + p_d \end{cases}$$

Best alignment

G	C	A	C	A	C	-	A
-	C	A	C	A	C	T	A



Complete example

Example: Local alignment of AGCACACA and ACACACTA

Costs: Match $m_{i,j} = +2$, Insertion/Déletion $p_i = p_j = -1$

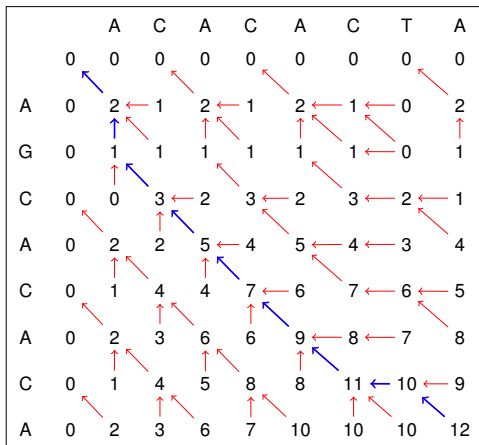
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Best alignment

A G C A C A C - A
A - C A C A C T A



Necessary properties:

- ▶ **Correctness:** \forall sub-problem, the computed value must indeed maximize the objective function .

Proofs usually inductive, and quite technical, but very systematic.

Desirable properties of DP schemes:

- ▶ **Completeness** of space of solutions **generated by** decomposition. Algorithmic tricks, by *cutting branches*, may violate this property.
- ▶ **Unambiguity:** Each solution is **generated** at most once.

⇒ Under these properties, one can **enumerate** solution space.

1 Introduction

- Dynamic programming 101
- Why RNA?
- RNA folding
- RNA Structure(s)
- Some representations of RNA structure

2 Some flavours of folding prediction

- Thermodynamics vs Kinetics
- Dynamic programming: Reminder

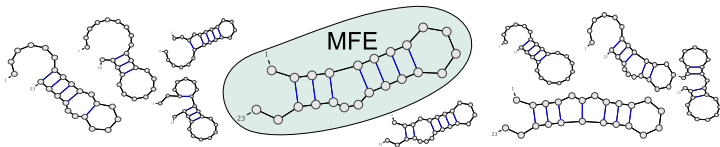
3 Free-energy minimization

- Nussinov-style RNA folding
- Turner energy model
- MFold/Unafold
- Performances and the comparative approach
- Towards a 3D ab-initio prediction

Problem A: Determine Minimum Free-Energy structure (MFE).

Ab initio folding prediction =

Predict RNA structure from its sequence ω only.



- ▶ **Conformations:** Set S_ω of secondary structures **compatible** (w.r.t. **base-pairing constraints**) with primary structure ω .
- ▶ **Free-Energy:** Function $E_{\omega,S}$ (KCal.mol⁻¹), **additive** on motifs occurring in any sequence/conformation couple (ω, S) .
- ▶ **Native structure:** Functional conformation of the biomolecule.

Remarks:

- ▶ Not necessarily unique (Kinetics, or bi-stable structures);
- ▶ In presence of PKs → Ambiguous: Which is the native conformation?

Nussinov/Jacobson energy model (NJ)

Base-pair maximization (with a twist):

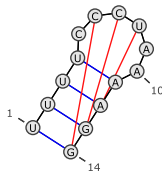
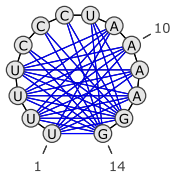
- ▶ Additive model on **independently contributing** base-pairs;
- ▶ **Canonical base-pairs** only: Watson/Crick (A/U,C/G) and Wobble (G/U)

$$\Rightarrow E_{\omega,S} = -\#Paires(S)$$

Folding in NJ model \Leftrightarrow **Base-pair** (weight) maximization

Example:

UUUUCCCUAAAAGG



Variant: Weight each pair with $-\#Hydrogen\ bonds$

$$\Delta G(G \equiv C) = -3$$

$$\Delta G(A = U) = -2$$

$$\Delta G(G - U) = -1$$

Nussinov/Jacobson energy model (NJ)

Base-pair maximization (with a twist):

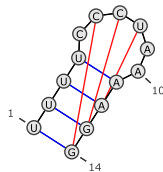
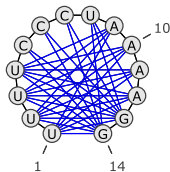
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Folding in NJ model \Leftrightarrow **Base-pair** (**weight**) maximization

Example:

UUUUCCCUAAAAGG



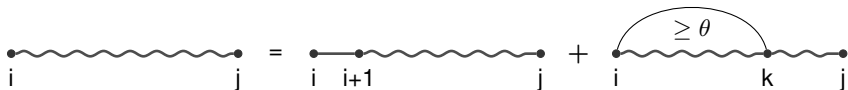
Variant: Weight each pair with $-\#Hydrogen\ bonds$

$$\Delta G(G \equiv C) = -3$$

$$\Delta G(A = U) = -2$$

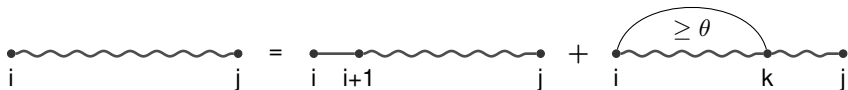
$$\Delta G(G - U) = -1$$

Nussinov/Jacobson DP scheme



$$N_{i,t} = 0, \quad \forall t \in [i, i + \theta]$$

$$N_{i,j} = \min \begin{cases} N_{i+1,j} & i \text{ unpaired} \\ \min_{k=i+\theta+1}^j \Delta G_{i,k} + N_{i+1,k-1} + N_{k+1,j} & i \text{ paired with } k \end{cases}$$

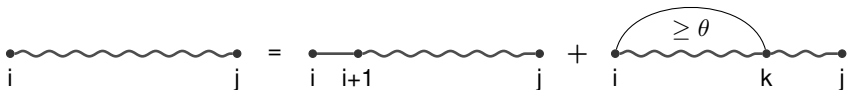


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Correctness. Goal = Show that MFE over interval $[i, j]$ is indeed found in $N_{i,j}$ after completing the computation. Proceed by induction:

- ▶ Assume that property holds for any $[i', j']$ such that $j' - i' < n$.
- ▶ Consider $[i, j], j - i = n$. Let $\text{MFE}_{i,j} :=$ Base-pairs of best struct. on $[i, j]$. Then first position i in $\text{MFE}_{i,j}$ is either:
 - ▶ **Unpaired:** $\text{MFE}_{i,j} = \text{MFE}_{i+1,j}$ → free-energy = $N_{i+1,j}$
 - ▶ **Paired to k :** $\text{MFE}_{i,j} = \{(i, k)\} \cup \text{MFE}_{i+1,k-1} \cup \text{MFE}_{k+1,j}$.
(Indeed, any BP between $[i+1, k-1]$ and $[k+1, j]$ would cross (i, k))
→ free-energy = $\Delta G_{i,k} + N_{i+1,k-1} + N_{k+1,j}$

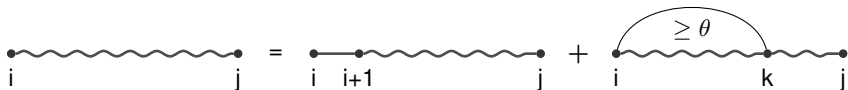


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$$N_{i,j} = \min \begin{cases} N_{i+1,j} & i \text{ unpaired} \\ \min_{k=i+\theta+1}^j \Delta G_{i,k} + N_{i+1,k-1} + N_{k+1,j} & i \text{ paired with } k \end{cases}$$

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 (Indeed, any BP between $[i + 1, k - 1]$ and $[k + 1, j]$ would cross (i, k))
 \rightarrow free-energy = $\Delta G_{i,k} + N_{i+1,k-1} + N_{k+1,j}$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10
A						0	0	0	0	0	2	2	2	5	5	5	8	8
C							0	0	0	0	0	0	2	5	5	5	8	8
U								0	0	0	0	0	2	3	5	5	6	7
U									0	0	0	0	2	3	5	5	5	7
C										0	0	0	0	3	3	3	5	5
U											0	0	0	0	2	2	2	3
U												0	0	0	0	0	1	2
A													0	0	0	0	0	0
G														0	0	0	0	0
A															0	0	0	0
C																0	0	0
G																	0	0
A																		0

$$\text{Sequence}(i, j) = \text{Sequence}(i+1, j) + \sum_{i \leq k < j} \text{Sequence}(i, k) + \text{Loop}(i, k, j)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10
A						0	0	0	0	0	2	2	2	5	5	5	8	8
C							0	0	0	0	0	0	2	5	5	5	8	8
U								0	0	0	0	0	2	3	5	5	6	7
U									0	0	0	0	2	3	5	5	5	7
C										0	0	0	0	3	3	3	5	5
U											0	0	0	0	2	2	2	3
U												0	0	0	0	0	1	2
A													0	0	0	0	0	0
G														0	0	0	0	0
A															0	0	0	0
C																0	0	0
G																	0	0
A																		0

$$\text{Seq}(i, j) = \text{Seq}(i+1, j) + \sum_{i \leq k < j, k-i \geq \theta} \text{Seq}(i, k) + \text{Seq}(k+1, j)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A	

C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14	
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11	
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10	
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10	
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10	
A						0	0	0	0	0	2	2	2	5	5	5	8	8	
C							0	0	0	0	0	0	2	5	5	5	8	8	
U								0	0	0	0	0	2	3	5	5	6	7	
U									0	0	0	0	2	3	5	5	5	7	
C										0	0	0	0	3	3	3	5	5	
U											0	0	0	0	2	2	2	3	
U												0	0	0	0	0	1	2	
A													0	0	0	0	0	0	
G														0	0	0	0	0	
A															0	0	0	0	
C																0	0	0	
G																	0	0	
A																		0	

$$\text{Sequence } i \dots j = \text{Sequence } i \dots i+1 + \text{Sequence } i \dots k + \text{Sequence } k \dots j$$

The second term in the sum is only valid if the distance between i and k is at least θ ($\geq \theta$).

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10
A						0	0	0	0	0	2	2	2	5	5	5	8	8
C							0	0	0	0	0	0	2	5	5	5	8	8
U								0	0	0	0	0	2	3	5	5	6	7
U									0	0	0	0	2	3	5	5	5	7
C										0	0	0	0	3	3	3	5	5
U											0	0	0	0	2	2	2	3
U												0	0	0	0	0	1	2
A													0	0	0	0	0	0
G														0	0	0	0	0
A															0	0	0	0
C																0	0	0
G																	0	0
A																		0

$$\text{Seq}(i, j) = \text{Seq}(i+1, j) + \sum_{i \leq k < j, \text{loop}(i, k) \geq \theta} \text{Seq}(i, k)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A	
	(.)	.
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14	
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11	
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10	
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10	
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10	
A						0	0	0	0	0	2	2	2	5	5	5	8	8	
C							0	0	0	0	0	0	2	5	5	5	8	8	
U								0	0	0	0	0	2	3	5	5	6	7	
U									0	0	0	0	2	3	5	5	5	7	
C										0	0	0	0	3	3	3	5	5	
U											0	0	0	0	2	2	2	3	
U												0	0	0	0	0	1	2	
A													0	0	0	0	0	0	
G														0	0	0	0	0	
A															0	0	0	0	
C																0	0	0	
G																	0	0	
A																		0	

$$\text{Sequence } i \dots j = \text{Sequence } i \dots i+1 + \text{Sequence } i \dots k + \text{Sequence } k \dots j \quad (\text{with } \geq \theta \text{ between } i \text{ and } k)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A
	(.)	.
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10
A						0	0	0	0	0	2	2	2	5	5	5	8	8
C							0	0	0	0	0	0	2	5	5	5	8	8
U								0	0	0	0	0	2	3	5	5	6	7
U									0	0	0	0	2	3	5	5	5	7
C										0	0	0	0	3	3	3	5	5
U											0	0	0	0	2	2	2	3
U												0	0	0	0	0	1	2
A													0	0	0	0	0	0
G														0	0	0	0	0
A															0	0	0	0
C																0	0	0
G																	0	0
A																		0

$$\text{Seq}(i, j) = \text{Seq}(i+1, j) + \max_{k \geq i+\theta} \{ \text{Seq}(i, k) + \text{Seq}(k+1, j) \}$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A	
	(.)	.
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14	
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11	
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10	
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10	
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10	
A						0	0	0	0	0	2	2	2	5	5	5	8	8	
C							0	0	0	0	0	0	2	5	5	5	8	8	
U								0	0	0	0	0	2	3	5	5	6	7	
U									0	0	0	0	2	3	5	5	5	7	
C										0	0	0	0	3	3	3	5	5	
U											0	0	0	0	2	2	2	3	
U												0	0	0	0	0	1	2	
A													0	0	0	0	0	0	
G														0	0	0	0	0	
A															0	0	0	0	
C																0	0	0	
G																	0	0	
A																		0	

$$\text{wavy}(i, j) = \text{wavy}(i, i+1) + \text{wavy}(i+1, j) + \sum_{k=i+1}^j \text{wavy}(i, k) + \text{wavy}(k, j) \quad \text{if } j-i \geq \theta$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A	
	(.)	.
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14	
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11	
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10	
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10	
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10	
A						0	0	0	0	0	2	2	2	5	5	5	8	8	
C							0	0	0	0	0	0	2	5	5	5	8	8	
U								0	0	0	0	0	2	3	5	5	6	7	
U									0	0	0	0	2	3	5	5	5	7	
C										0	0	0	0	3	3	3	5	5	
U											0	0	0	0	2	2	2	3	
U												0	0	0	0	0	1	2	
A													0	0	0	0	0	0	
G														0	0	0	0	0	
A															0	0	0	0	
C																0	0	0	
G																	0	0	
A																		0	

$$\text{Teal}(i, j) = \text{Purple}(i, i+1) + \text{Red}(i, k) + \text{Blue}(i, k) \text{ where } \theta \leq \text{arc}(i, k)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A	
	(.)	.
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14	
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11	
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10	
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10	
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10	
A						0	0	0	0	0	2	2	2	5	5	5	8	8	
C							0	0	0	0	0	0	2	5	5	5	8	8	
U								0	0	0	0	0	2	3	5	5	6	7	
U									0	0	0	0	2	3	5	5	5	7	
C										0	0	0	0	3	3	3	5	5	
U											0	0	0	0	2	2	2	3	
U												0	0	0	0	0	1	2	
A													0	0	0	0	0	0	
G														0	0	0	0	0	
A															0	0	0	0	
C																0	0	0	
G																	0	0	
A																		0	

$$\text{Sequence}(i, j) = \text{Sequence}(i, i+1) + \text{Sequence}(i, k) + \text{Sequence}(k, j) \quad (\text{with } \geq \theta)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A	
	(.)	.
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14	
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11	
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10	
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10	
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10	
A						0	0	0	0	0	2	2	2	5	5	5	8	8	
C							0	0	0	0	0	0	2	5	5	5	8	8	
U								0	0	0	0	0	2	3	5	5	6	7	
U									0	0	0	0	2	3	5	5	5	7	
C										0	0	0	0	3	3	3	5	5	
U											0	0	0	0	2	2	2	3	
U												0	0	0	0	0	1	2	
A													0	0	0	0	0	0	
G														0	0	0	0	0	
A															0	0	0	0	
C																0	0	0	
G																	0	0	
A																		0	

$$\text{wavy}(i, j) = \text{wavy}(i, i+1) + \text{wavy}(i, k) + \text{wavy}(k, j)$$

The diagram shows a wavy line from index i to j (teal) is equal to the sum of:

- A wavy line from i to $i+1$ (purple).
- A wavy line from i to k (orange) with a loop of size $\geq \theta$ connecting i and k .
- A wavy line from k to j (red).

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A	
	(.)	.
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14	
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11	
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10	
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10	
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10	
A						0	0	0	0	0	2	2	2	5	5	5	8	8	
C							0	0	0	0	0	0	2	5	5	5	8	8	
U								0	0	0	0	0	2	3	5	5	6	7	
U									0	0	0	0	2	3	5	5	5	7	
C										0	0	0	0	3	3	3	5	5	
U											0	0	0	0	2	2	2	3	
U												0	0	0	0	0	1	2	
A													0	0	0	0	0	0	
G														0	0	0	0	0	
A															0	0	0	0	
C																0	0	0	
G																	0	0	
A																		0	

$i \text{---} j = i \text{---} i+1 \text{---} j + i \text{---} k \text{---} j$

 $k \geq i+1$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A	
	(.)	.
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14	
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11	
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10	
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10	
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10	
A						0	0	0	0	0	2	2	2	5	5	5	8	8	
C							0	0	0	0	0	0	2	5	5	5	8	8	
U								0	0	0	0	0	2	3	5	5	6	7	
U									0	0	0	0	2	3	5	5	5	7	
C										0	0	0	0	3	3	3	5	5	
U											0	0	0	0	2	2	2	3	
U												0	0	0	0	0	1	2	
A													0	0	0	0	0	0	
G														0	0	0	0	0	
A															0	0	0	0	
C																0	0	0	
G																	0	0	
A																		0	

$$\text{Sequence } i \dots j = \text{Sequence } i \dots i+1 + \text{Sequence } i \dots k + \text{Sequence } k \dots j$$

The second term in the sum is associated with the inequality $\geq \theta$.

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A
	(.)	.
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10
A						0	0	0	0	0	2	2	2	5	5	5	8	8
C							0	0	0	0	0	0	2	5	5	5	8	8
U								0	0	0	0	0	2	3	5	5	6	7
U									0	0	0	0	2	3	5	5	5	7
C										0	0	0	0	3	3	3	5	5
U											0	0	0	0	2	2	2	3
U												0	0	0	0	0	1	2
A													0	0	0	0	0	0
G														0	0	0	0	0
A															0	0	0	0
C																0	0	0
G																	0	0
A																		0

$$\text{wavy}(i, j) = \text{wavy}(i, i+1) + \text{wavy}(i+1, j) + \sum_{i < k < j} \text{wavy}(i, k) + \text{loop}(i, k) + \text{wavy}(k, j)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A	
	((.))	.	
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14	
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11	
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10	
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10	
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10	
A						0	0	0	0	0	2	2	2	5	5	5	8	8	
C							0	0	0	0	0	0	2	5	5	5	8	8	
U								0	0	0	0	0	2	3	5	5	6	7	
U									0	0	0	0	2	3	5	5	5	7	
C										0	0	0	0	3	3	3	5	5	
U											0	0	0	0	2	2	2	3	
U												0	0	0	0	0	1	2	
A													0	0	0	0	0	0	
G														0	0	0	0	0	
A															0	0	0	0	
C																0	0	0	
G																	0	0	
A																		0	

$$\text{Seq}(i, j) = \text{Seq}(i+1, j) + \sum_{k=i+\theta}^j \text{Seq}(i, k)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A
	((.))	.
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10
A						0	0	0	0	0	2	2	2	5	5	5	8	8
C							0	0	0	0	0	0	2	5	5	5	8	8
U								0	0	0	0	0	2	3	5	5	6	7
U									0	0	0	0	2	3	5	5	5	7
C										0	0	0	0	3	3	3	5	5
U											0	0	0	0	2	2	2	3
U												0	0	0	0	0	1	2
A													0	0	0	0	0	0
G														0	0	0	0	0
A															0	0	0	0
C																0	0	0
G																	0	0
A																		0

$$\text{Seq}(i, j) = \text{Seq}(i+1, j) + \sum_{k=i+\theta}^j \text{Seq}(i, k) + \text{Loop}(i, k, j)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A
	((.))	.
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10
A						0	0	0	0	0	2	2	2	5	5	5	8	8
C							0	0	0	0	0	0	2	5	5	5	8	8
U								0	0	0	0	0	2	3	5	5	6	7
U									0	0	0	0	2	3	5	5	5	7
C										0	0	0	0	3	3	3	5	5
U											0	0	0	0	2	2	2	3
U												0	0	0	0	0	1	2
A													0	0	0	0	0	0
G														0	0	0	0	0
A															0	0	0	0
C																0	0	0
G																	0	0
A																		0

$$\text{Seq}(i, j) = \text{Seq}(i+1, j) + \sum_{i \leq k < j, k-i \geq \theta} \text{Seq}(i, k)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A
	(((.	.	.)))	.
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10
A						0	0	0	0	0	2	2	2	5	5	5	8	8
C							0	0	0	0	0	0	2	5	5	5	8	8
U								0	0	0	0	0	2	3	5	5	6	7
U									0	0	0	0	2	3	5	5	5	7
C										0	0	0	0	3	3	3	5	5
U											0	0	0	0	2	2	2	3
U												0	0	0	0	0	1	2
A													0	0	0	0	0	0
G														0	0	0	0	0
A															0	0	0	0
C																0	0	0
G																	0	0
A																		0

$$\text{Sequence } i \dots j = \text{Sequence } i \dots i+1 + \text{Sequence } i \dots k + \text{Sequence } k \dots j \quad (\text{with } \geq \theta \text{ above } i \dots k)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A
	(((.	.	.)))	.
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10
A						0	0	0	0	0	2	2	2	5	5	5	8	8
C							0	0	0	0	0	0	2	5	5	5	8	8
U								0	0	0	0	0	2	3	5	5	6	7
U									0	0	0	0	2	3	5	5	5	7
C										0	0	0	0	3	3	3	5	5
U											0	0	0	0	2	2	2	3
U												0	0	0	0	0	1	2
A													0	0	0	0	0	0
G														0	0	0	0	0
A															0	0	0	0
C																0	0	0
G																	0	0
A																		0

$$\text{Sequence } i \dots j = \text{Sequence } i \dots i+1 + \text{Sequence } i \dots k + \text{Sequence } k \dots j \quad (\text{with } \geq \theta \text{ constraint})$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A	
	(((.	.	.)))	.	
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14	
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11	
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10	
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10	
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10	
A						0	0	0	0	0	2	2	2	5	5	5	8	8	
C							0	0	0	0	0	0	2	5	5	5	8	8	
U								0	0	0	0	0	2	3	5	5	6	7	
U									0	0	0	0	2	3	5	5	5	7	
C										0	0	0	0	3	3	3	5	5	
U											0	0	0	0	2	2	2	3	
U												0	0	0	0	0	1	2	
A													0	0	0	0	0	0	
G														0	0	0	0	0	
A															0	0	0	0	
C																0	0	0	
G																	0	0	
A																		0	

$$\text{Sequence } i \dots j = \text{Sequence } i \dots i+1 + \text{Sequence } i \dots k + \text{Sequence } k \dots j \quad (\text{with } \geq \theta \text{ between } i \text{ and } k)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A
	(((.	.	.)))	.
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10
A						0	0	0	0	0	2	2	2	5	5	5	8	8
C							0	0	0	0	0	0	2	5	5	5	8	8
U								0	0	0	0	0	2	3	5	5	6	7
U									0	0	0	0	2	3	5	5	5	7
C										0	0	0	0	3	3	3	5	5
U											0	0	0	0	2	2	2	3
U												0	0	0	0	0	1	2
A													0	0	0	0	0	0
G														0	0	0	0	0
A															0	0	0	0
C																0	0	0
G																	0	0
A																		0

$$\text{Sequence}(i, j) = \text{Sequence}(i, i+1) + \text{Sequence}(i+1, j) + \sum_{k=i+1}^j \text{Sequence}(i, k) + \text{Sequence}(k, j) \quad (\text{with } \theta \leq k-i)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A
	(((.	.	.)))	.
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10
A						0	0	0	0	0	2	2	2	5	5	5	8	8
C							0	0	0	0	0	0	2	5	5	5	8	8
U								0	0	0	0	0	2	3	5	5	6	7
U									0	0	0	0	2	3	5	5	5	7
C										0	0	0	0	3	3	3	5	5
U											0	0	0	0	2	2	2	3
U												0	0	0	0	0	1	2
A													0	0	0	0	0	0
G														0	0	0	0	0
A															0	0	0	0
C																0	0	0
G																	0	0
A																		0

$$i \text{---} j = i \text{---} i+1 \text{---} j + i \text{---} k \text{---} j \quad (\text{with } \geq \theta \text{ between } i \text{ and } k)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A
	(((.	.	.)))	.
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10
A						0	0	0	0	0	2	2	2	5	5	5	8	8
C							0	0	0	0	0	0	2	5	5	5	8	8
U								0	0	0	0	0	2	3	5	5	6	7
U									0	0	0	0	2	3	5	5	5	7
C										0	0	0	0	3	3	3	5	5
U											0	0	0	0	2	2	2	3
U												0	0	0	0	0	1	2
A													0	0	0	0	0	0
G														0	0	0	0	0
A															0	0	0	0
C																0	0	0
G																	0	0
A																		0

$$\text{Sequence } i \dots j = \text{Sequence } i \dots i+1 + \text{Sequence } i \dots k + \text{Sequence } k \dots j \quad (\text{with } \geq \theta \text{ between } i \text{ and } k)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A
	(((.	.	.)))	.
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10
A						0	0	0	0	0	2	2	2	5	5	5	8	8
C							0	0	0	0	0	0	2	5	5	5	8	8
U								0	0	0	0	0	2	3	5	5	6	7
U									0	0	0	0	2	3	5	5	5	7
C										0	0	0	0	3	3	3	5	5
U											0	0	0	0	2	2	2	3
U												0	0	0	0	0	1	2
A													0	0	0	0	0	0
G														0	0	0	0	0
A															0	0	0	0
C																0	0	0
G																	0	0
A																		0

$$\text{wavy}(i, j) = \text{wavy}(i, i+1) + \text{wavy}(i+1, j) + \text{wavy}(i, k) + \text{wavy}(k, j) \quad (\text{with } \theta \text{ pairing between } i \text{ and } k)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A	
	(((.	.	.)	.	(.)))	.	
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14	
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11	
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10	
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10	
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10	
A						0	0	0	0	0	2	2	2	5	5	5	8	8	
C							0	0	0	0	0	0	2	5	5	5	8	8	
U								0	0	0	0	0	2	3	5	5	6	7	
U									0	0	0	0	2	3	5	5	5	7	
C										0	0	0	0	3	3	3	5	5	
U											0	0	0	0	2	2	2	3	
U												0	0	0	0	0	1	2	
A													0	0	0	0	0	0	
G														0	0	0	0	0	
A															0	0	0	0	
C																0	0	0	
G																	0	0	
A																		0	

$$\text{wavy}(i, j) = \text{wavy}(i, i+1) + \text{wavy}(i+1, j) + \text{wavy}(i, k) + \text{wavy}(k, j) \quad (\text{with } \geq \theta)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A	
	(((.	.	.)	.	(.)))	.	
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14	
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11	
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10	
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10	
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10	
A						0	0	0	0	0	2	2	2	5	5	5	8	8	
C							0	0	0	0	0	0	2	5	5	5	8	8	
U								0	0	0	0	0	2	3	5	5	6	7	
U									0	0	0	0	2	3	5	5	5	7	
C										0	0	0	0	3	3	3	5	5	
U											0	0	0	0	2	2	2	3	
U												0	0	0	0	0	1	2	
A													0	0	0	0	0	0	
G														0	0	0	0	0	
A															0	0	0	0	
C																0	0	0	
G																	0	0	
A																		0	

$$\text{Seq}(i, j) = \text{Seq}(i+1, j) + \sum_{i \leq k < j, k-i \geq \theta} \text{Seq}(i, k)$$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A	
	(((.	.	.)	.	(.)))	.	
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14	
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11	
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10	
A				0	0	0	2	2	2	2	4	4	5	7	7	8	10	10	
U					0	0	0	0	0	2	2	4	5	7	7	8	10	10	
A						0	0	0	0	2	2	2	5	5	5	8	8	8	
C							0	0	0	0	0	2	5	5	5	8	8	8	
U								0	0	0	0	2	3	5	5	6	7	7	
U									0	0	0	2	3	5	5	5	7	7	
C										0	0	0	3	3	3	5	5	5	
U											0	0	2	2	2	2	3	3	
U												0	0	0	0	1	2	2	
A													0	0	0	0	0	0	
G														0	0	0	0	0	
A															0	0	0	0	
C																0	0	0	
G																	0	0	
A																		0	

$i \text{---} j = i \text{---} i+1 \text{---} j + i \text{---} k \text{---} j$

$k \geq i+1$

	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A	
	(((.	.	.)	.	((.	.	.))))	.	
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14	
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11	
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10	
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10	
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10	
A						0	0	0	0	0	2	2	2	5	5	5	8	8	
C							0	0	0	0	0	0	2	5	5	5	8	8	
U								0	0	0	0	0	2	3	5	5	6	7	
U									0	0	0	0	2	3	5	5	5	7	
C										0	0	0	0	3	3	3	5	5	
U											0	0	0	0	2	2	2	3	
U												0	0	0	0	0	1	2	
A													0	0	0	0	0	0	
G														0	0	0	0	0	
A															0	0	0	0	
C																0	0	0	
G																	0	0	
A																		0	

$$\text{Sequence}(i, j) = \text{Sequence}(i, i+1) + \text{Sequence}(i, k) + \text{Sequence}(k, j) \quad (\text{with } k - i \geq \theta)$$

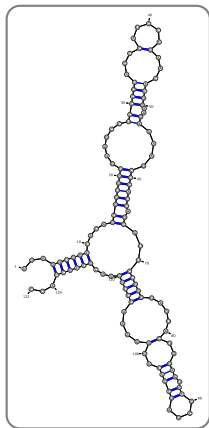
	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A	
	(((.	.	.)	.	((.	.	.))))	.	
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14	
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11	
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10	
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10	
U					0	0	0	0	0	0	2	2	4	5	7	7	8	10	
A						0	0	0	0	0	2	2	2	5	5	5	8	8	
C							0	0	0	0	0	0	2	5	5	5	8	8	
U								0	0	0	0	0	2	3	5	5	6	7	
U									0	0	0	0	2	3	5	5	5	7	
C										0	0	0	0	3	3	3	5	5	
U											0	0	0	0	2	2	2	3	
U												0	0	0	0	0	1	2	
A													0	0	0	0	0	0	
G														0	0	0	0	0	
A															0	0	0	0	
C																0	0	0	
G																	0	0	
A																		0	

$i \dots j = i \dots i+1 \dots j + i \dots k \dots j$

The second term includes a pairing arc between i and k with the condition $\geq \theta$.

Based on **unambiguous** decomposition of 2^{ary} structure into **loops**:

- ▶ Internal loops
- ▶ Bulges
- ▶ Terminal loops
- ▶ Multi loops
- ▶ Stackings



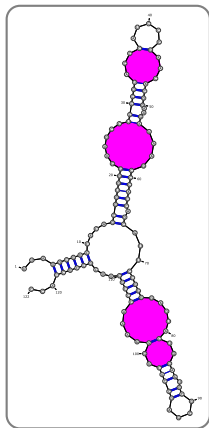
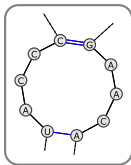
Free-energy ΔG of a loop depend on bases, assymetry, dangles ...

Experimentally determined
+ Interpolated for larger loops.

Improved results by taking stacking into account.

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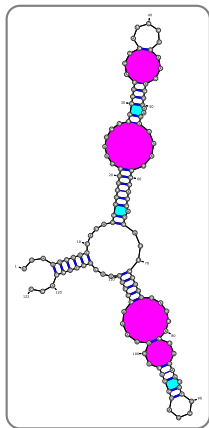
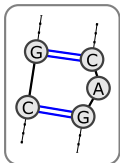
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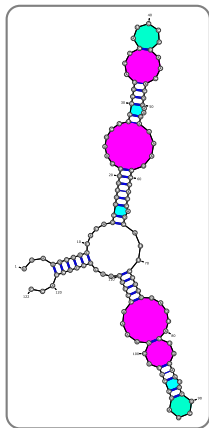
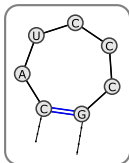
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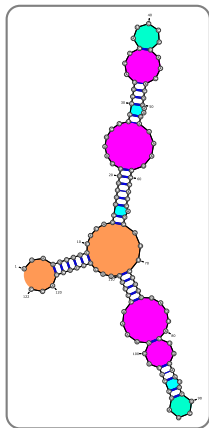
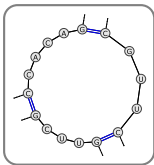
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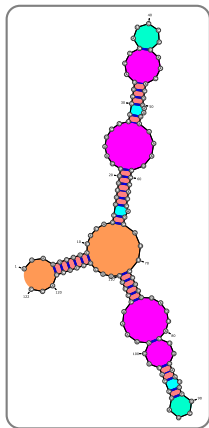
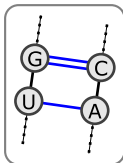
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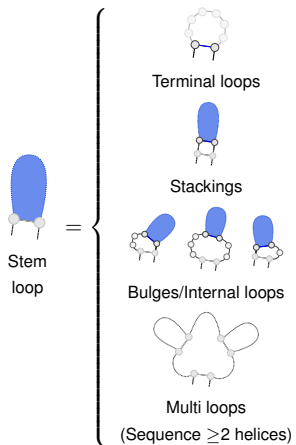


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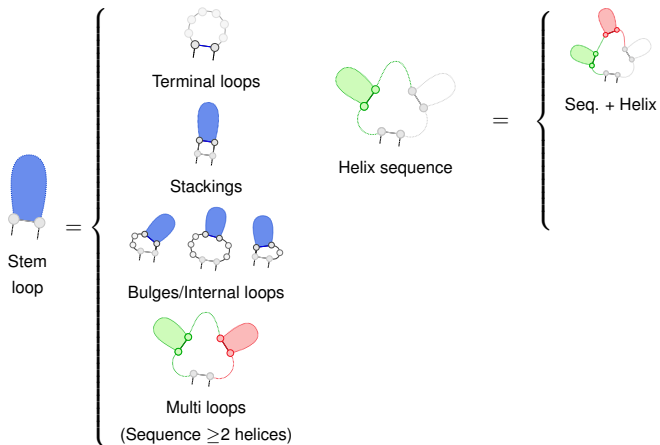
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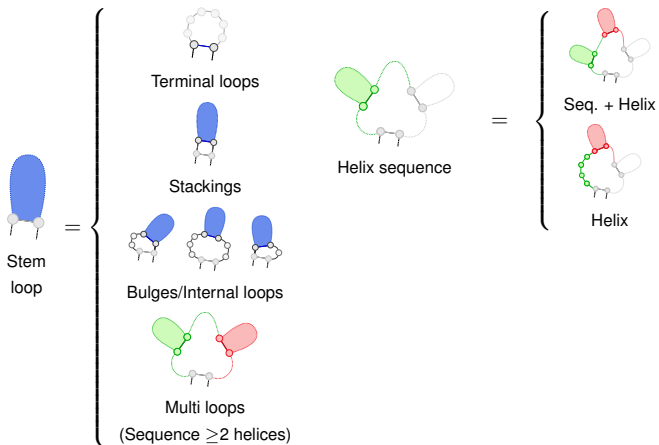
MFE DP equations



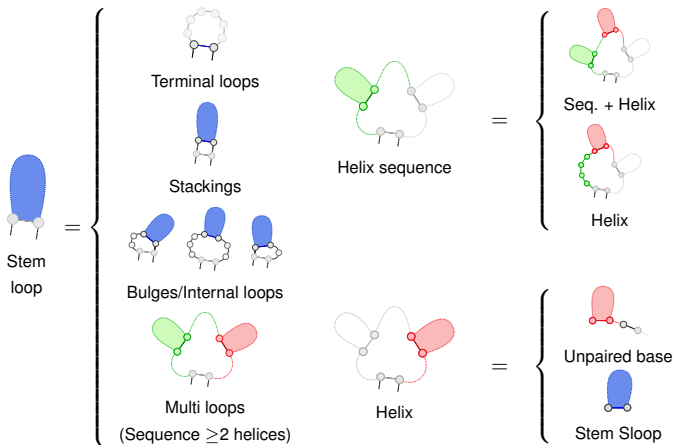
MFE DP equations



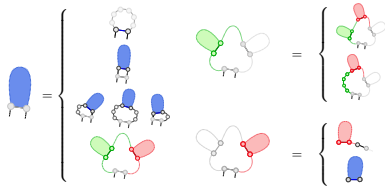
MFE DP equations



MFE DP equations



- ▶ $E_H(i, j)$: Energy of terminal loop *enclosed by* (i, j) pair
- ▶ $E_{BI}(i, j)$: Energy of bulge or internal loop *enclosed by* (i, j) pair
- ▶ $E_S(i, j)$: Energy of stacking $(i, j)/(i + 1, j - 1)$
- ▶ Penalty for multi loop (a), and occurrences of unpaired base (b) and helix (c) in multi loops.



DP recurrence

$$\begin{aligned}
 \mathcal{M}'_{i,j} &= \min \begin{cases} E_H(i, j) \\ E_S(i, j) + \mathcal{M}'_{i+1, j-1} \\ \text{Min}_{i', j'} (E_{BI}(i, i', j', j) + \mathcal{M}'_{i', j'}) \\ a + c + \text{Min}_k (\mathcal{M}_{i+1, k-1} + \mathcal{M}^1_{k, j-1}) \end{cases} \\
 \mathcal{M}_{i,j} &= \text{Min}_k \{ \min (\mathcal{M}_{i, k-1}, b(k-1)) + \mathcal{M}^1_{k, j} \} \\
 \mathcal{M}^1_{i,j} &= \text{Min}_k \{ b + \mathcal{M}^1_{i, j-1}, c + \mathcal{M}'_{i, j} \}
 \end{aligned}$$

Backtracking to reconstruct MFE structure:

$$\mathcal{M}'_{i,j} = \text{Min} \left\{ \begin{array}{l} E_H(i,j) \\ E_S(i,j) + \mathcal{M}'_{i+1,j-1} \\ \text{Min}_{i',j'} (E_{BI}(i,i',j',j) + \mathcal{M}'_{i',j'}) \\ a + c + \text{Min}_k (\mathcal{M}_{i+1,k-1} + \mathcal{M}^1_{k,j-1}) \end{array} \right\}$$
$$\mathcal{M}_{i,j} = \text{Min}_k \left\{ \min (\mathcal{M}_{i,k-1}, b(k-1)) + \mathcal{M}^1_{k,j} \right\}$$
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Complexity:

For each min, $\mathcal{O}(n)$ potential contributors

⇒ **Worst-case** complexity in $\mathcal{O}(n^2)$ for **naive backtrack**.

Keep best contributor for each Min ⇒ **Backtracking in $\mathcal{O}(n)$**

⇒ Unafold [?]/RNAfold [?] compute the MFE for the Turner model
in **overall**³ time/space complexities in $\mathcal{O}(n^3)/\mathcal{O}(n^2)$

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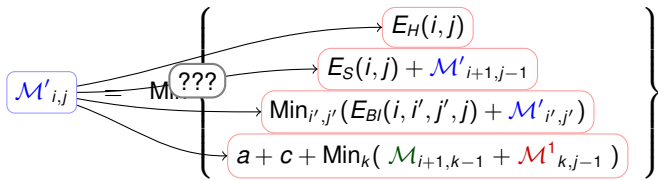
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Definition (Ab initio folding)

Starting from sequence, find conformation that minimizes free-energy.

Advantages:

- ▶ Mechanical nature allows the (in)validation of models
- ▶ Reasonable complexity $\mathcal{O}(n^3)/\mathcal{O}(n^2)$ time/space
- ▶ *Exhaustive* nature

Limitations:

- ▶ Hard to include PKs
- ▶ Highly dependent on energy model
- ▶ No cooperativity
- ▶ Limited performances

Definition (Comparative approach)

Starting from homologous sequences, postulate common structure and find best possible tradeoff between folding & alignment.

Advantages :

- ▶ Better performances
- ▶ (Limited) cooperativity
- ▶ Self-improving

Limitations

- ▶ Easily unreasonable complexity
- ▶ Non exhaustive search
- ▶ Captures *transient* structures

Definition (Ab initio folding)

Starting from sequence, find conformation that minimizes free-energy.

Advantages:

- ▶ Mechanical nature allows the (in)validation of models
- ▶ Reasonable complexity $\mathcal{O}(n^3)/\mathcal{O}(n^2)$ time/space
- ▶ *Exhaustive* nature

Limitations:

- ▶ Hard to include PKs
- ▶ Highly dependent on energy model
- ▶ No cooperativity
- ▶ Limited performances

Definition (Comparative approach)

Starting from homologous sequences, postulate common structure and find best possible tradeoff between folding & alignment.

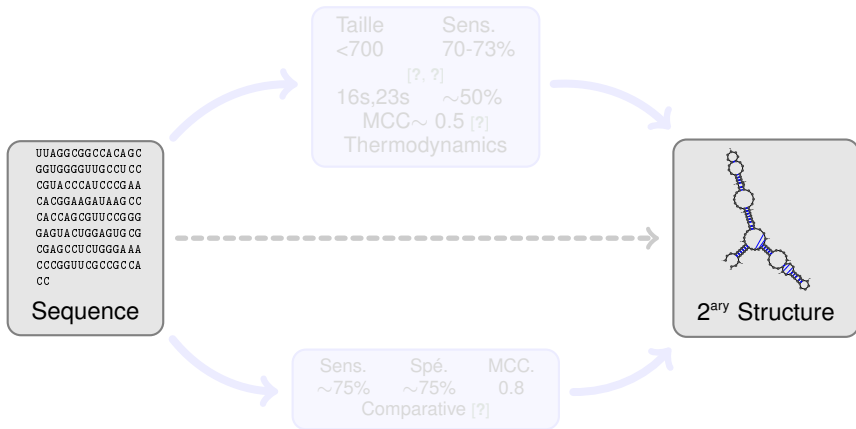
Avantages :

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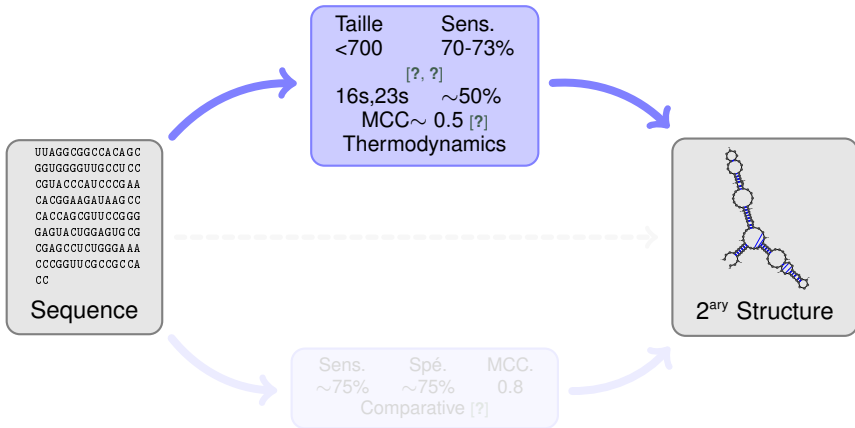
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- ▶ Easily unreasonable complexity
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Performances

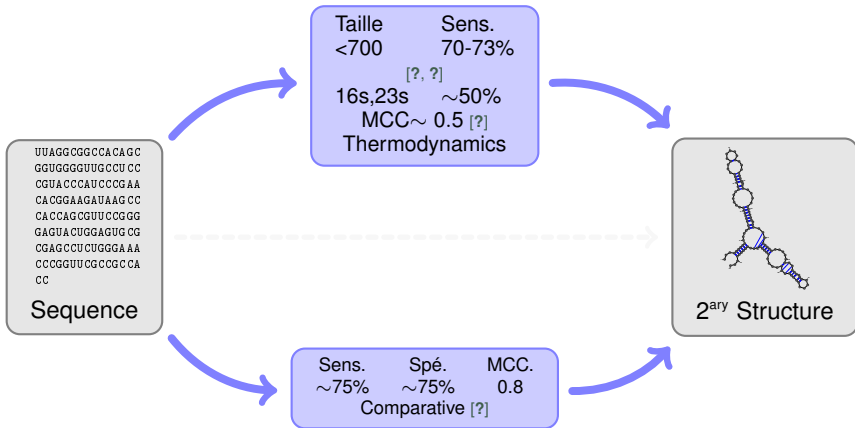


Reminder:
$$MCC = \frac{t^+t^- - t^+f^-}{\sqrt{(t^++f^+)(t^++f^-)(t^-+f^+)(t^-+f^-)}}$$



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Performances



Reminder: $MCC = \frac{t^+t^- - f^+f^-}{\sqrt{(t^++f^+)(t^++f^-)(t^-+f^+)(t^-+f^-)}}$

Goal: From sequence to all-atom/coarse grain 3D models!!!

- ▶ Comparative models + Molecular dynamics: RNA2D3D [?]
- ▶ Pipeline MC-Fold/MC-sym [?]

```
UUAGGCGGCCACAGC
GGUGGGGUUGCCUCC
CGUACCCAUCCGAA
CACGGAAGUAAGCC
CACCCAGCGUUCGGG
GAGUACUGGAGUGCG
CGAGCCUCUGGGA AA
CCCGGUUCGCCGCCA
CC
```

Séquence

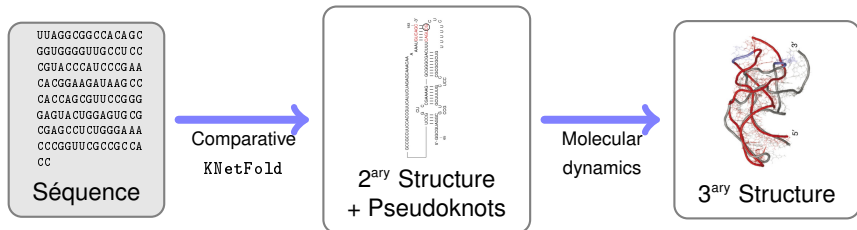


3^{ary} Structure

Towards a 3D ab-initio prediction

Goal: From sequence to all-atom/coarse grain 3D models!!!

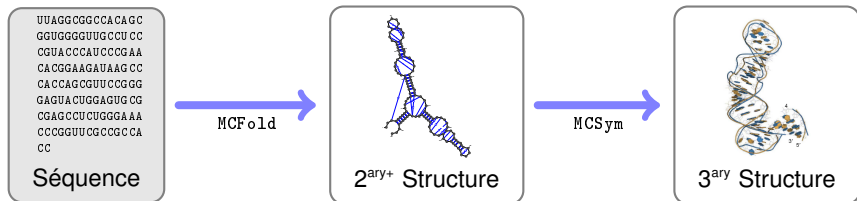
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Towards a 3D ab-initio prediction

Goal: From sequence to all-atom/coarse grain 3D models!!!

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Exercise: Parsing/folding RNAs (Python)

[http://www.lix.polytechnique.fr/~ponty/enseignement/
2015-2016-BIM-TP1.pdf](http://www.lix.polytechnique.fr/~ponty/enseignement/2015-2016-BIM-TP1.pdf)