

# M2 BIM – STRUCT – Lecture 1

## Folding RNA *in silico*

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AMIBio Team  
CNRS LIX, École Polytechnique

# Outline

## Introduction

Dynamic programming 101

Why RNA?

RNA folding

RNA Structure(s)

Some representations of RNA structure

## Some flavours of folding prediction

Thermodynamics vs Kinetics

Dynamic programming: Reminder

## Free-energy minimization

Nussinov-style RNA folding

## Foreword...

...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 =??

55

60

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60??

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$$21 = \text{50c} + \text{1c}$$

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$$60 = \text{50c} + \text{1c} + \text{1c} + \text{1c} + \text{1c} + \text{1c} + \text{1c} + \text{1c} + \text{1c} + \text{1c} + \text{1c} \quad ??$$

$$= \text{20c} + \text{20c} + \text{20c} \quad !$$

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

## Foreword...

**Strategy #2:** Brute force enumeration  $\rightarrow$  #Coins<sup>N</sup> (Ouch!)

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

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

With some memory ( $N$  intermediate computations), the minimum number of coins can be obtained after  $N \times \text{\#Coins}$  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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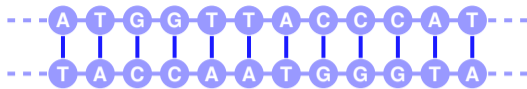
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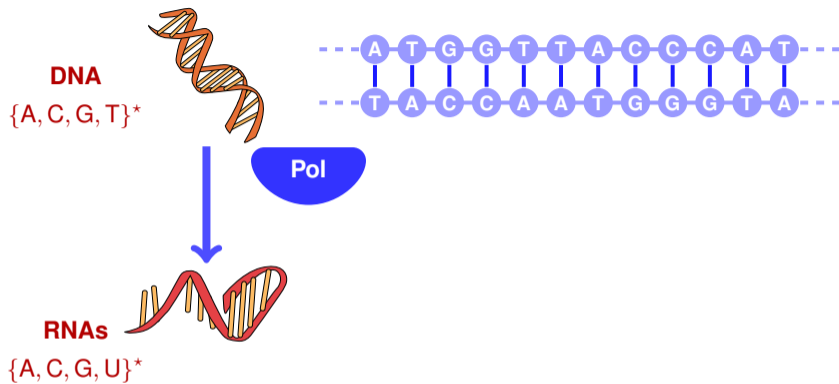
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# Fundamental *dogma* of molecular biology

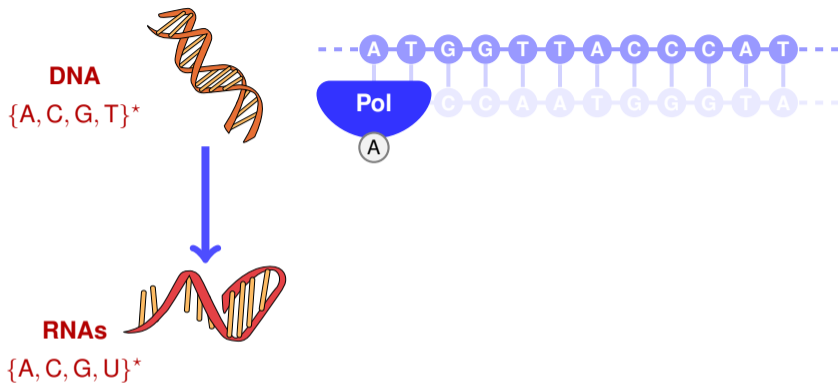
**DNA**  
{A, C, G, T}<sup>\*</sup>



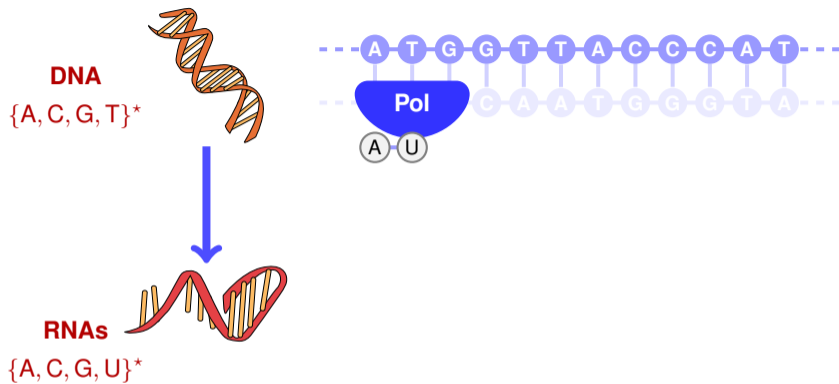
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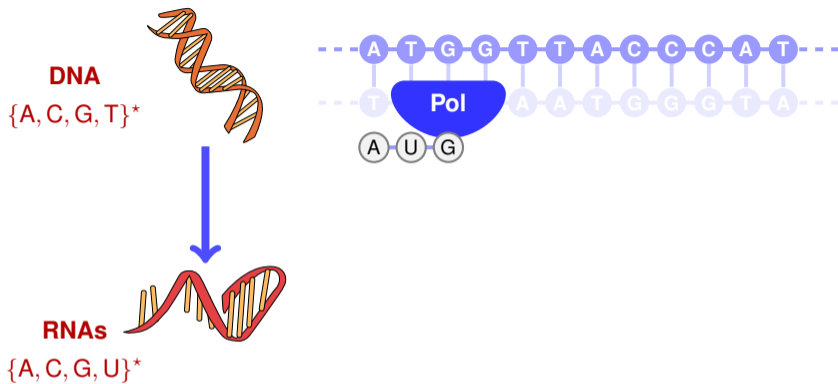
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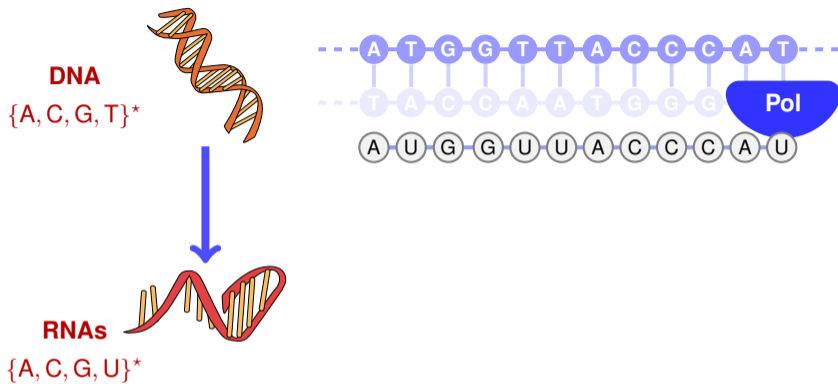
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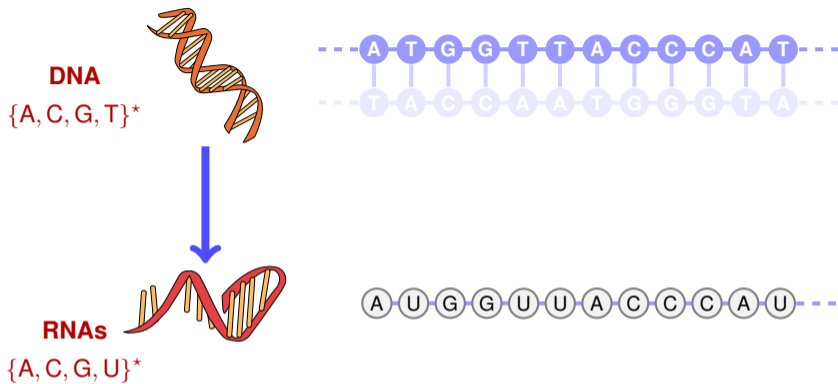
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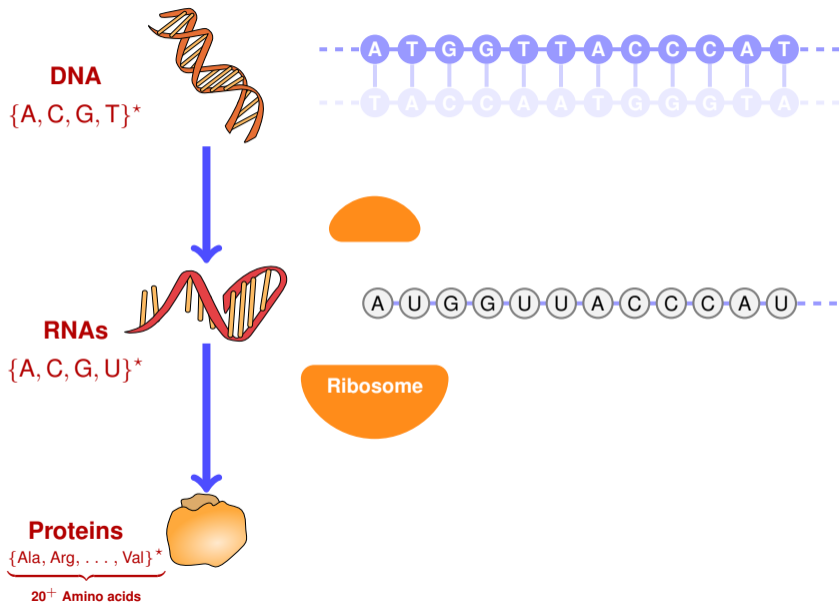


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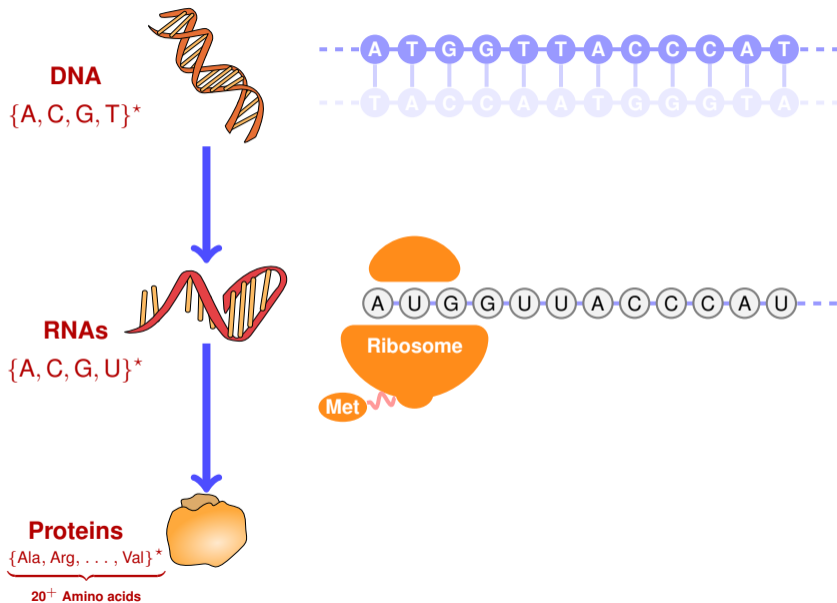




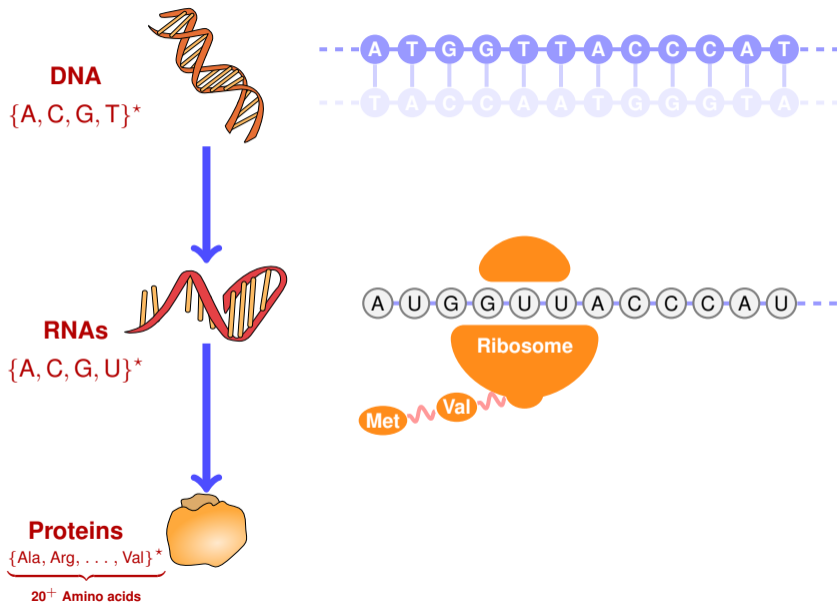
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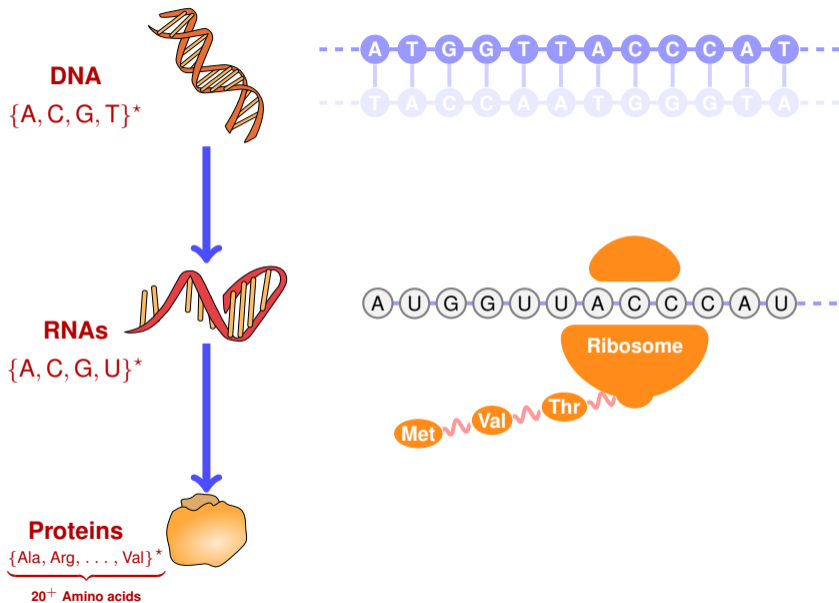
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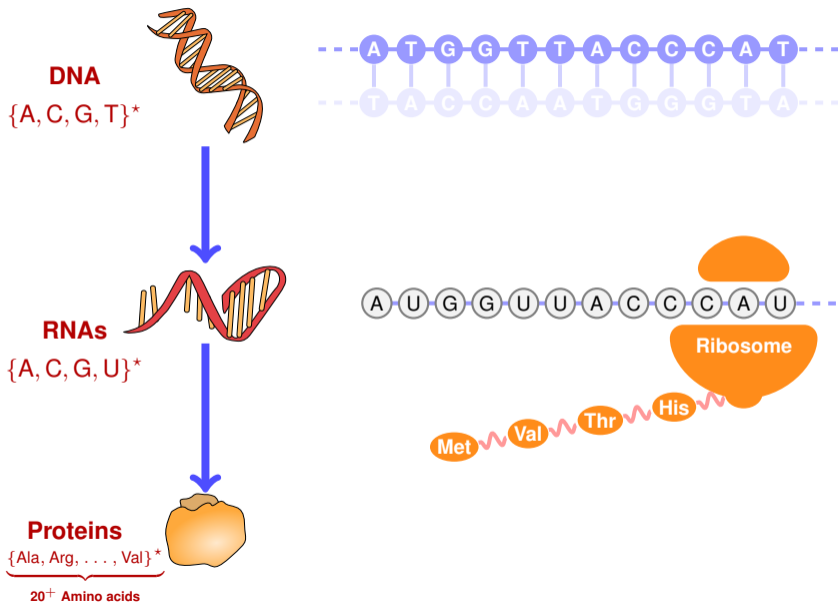
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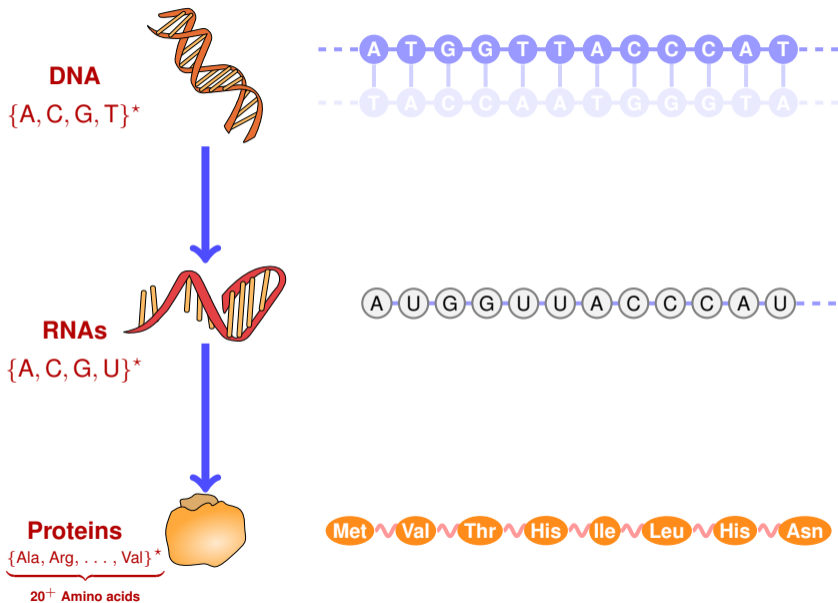
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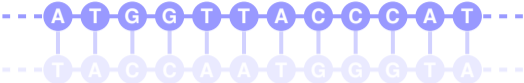


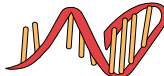
# Fundamental *dogma* of molecular biology




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**THE CODE**  
(genes)  
**DNA**  
{A, C, G, T}\*  




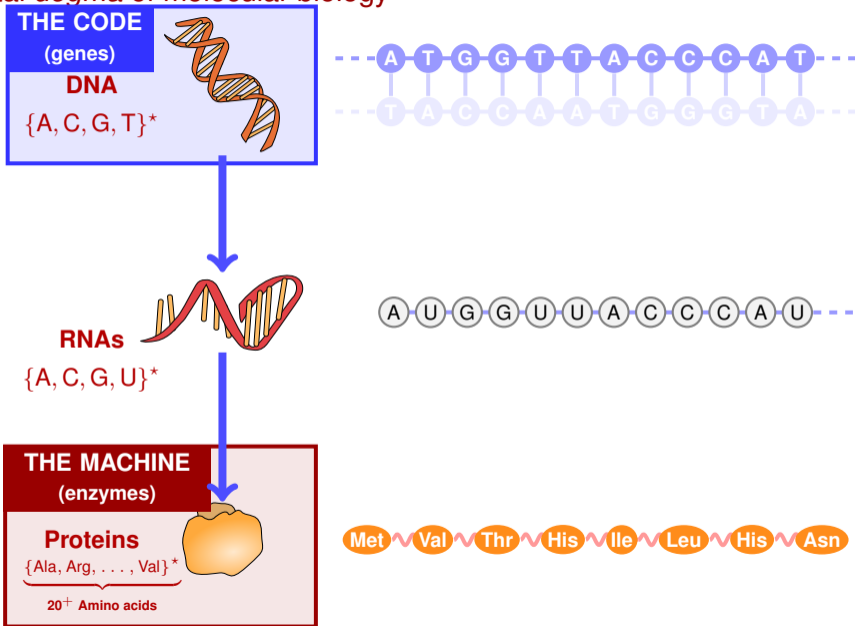
**RNAs**  
{A, C, G, U}\*  




**Proteins**  
{Ala, Arg, . . . , Val}\*  
  
20+ Amino acids



# Fundamental *dogma* of molecular biology

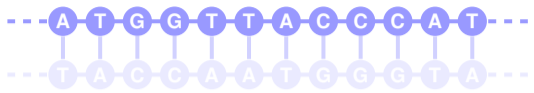




# Fundamental *dogma* of molecular biology

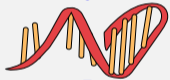
**THE CODE**  
(genes)

**DNA**  
{A, C, G, T}<sup>\*</sup>



**MEH...**


**RNAs**  
{A, C, G, U}<sup>\*</sup>



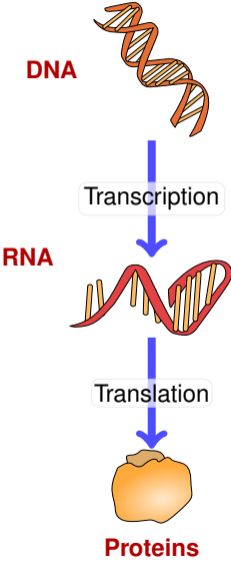
**THE MACHINE**  
(enzymes)

**Proteins**  
{Ala, Arg, ..., Val}<sup>\*</sup>

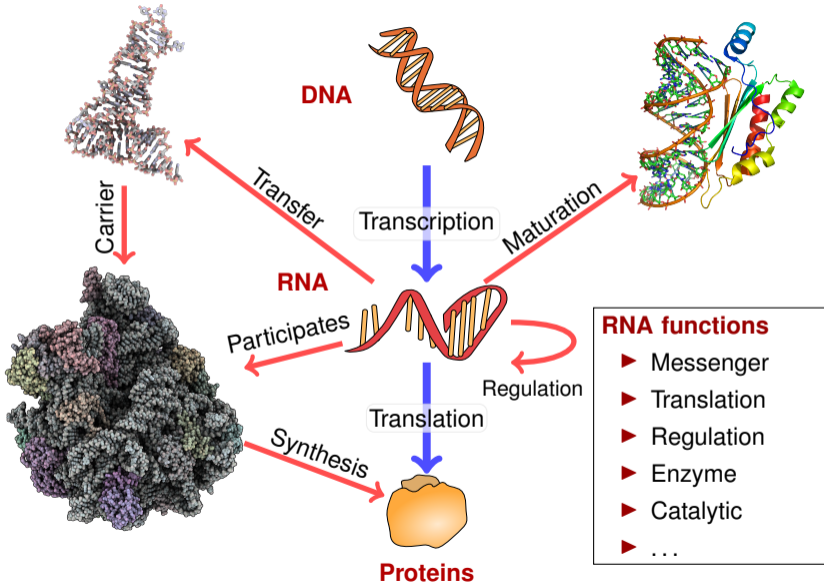
20<sup>+</sup> Amino acids



# Fundamental *dogma* of molecular biology



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# RiboNucleic Acids (RNAs) in Human biology/health: Friends **and** Foes!

## RiboNucleic Acids (RNAs)

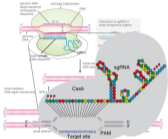


**Encodes proteins**  
mRNA Vaccines  
COVID-19, Malaria (Zika, CMV, Cancers?)

# RiboNucleic Acids (RNAs) in Human biology/health: Friends **and** Foes!

## Targeting system for DNA Editing

CRISPR therapies  
Sickle-cell anemia,  $\beta$ -thalassamia, Leber congenital amaurosis (LCA), cancers...



Hendel et al, 2015; Agrotis & Ketteler, 2015

## RiboNucleic Acids (RNAs)



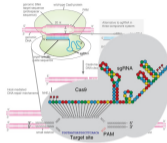
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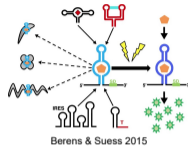
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Berens & Süss 2015

## Sensor of metabolites

Riboswitches

## RiboNucleic Acids (RNAs)



## Encodes proteins

mRNA Vaccines

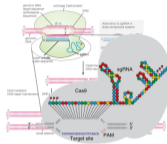
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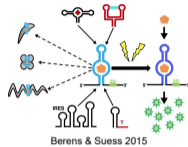
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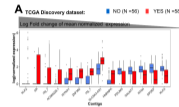
Berens & Suess 2015

Sensor of metabolites  
Riboswitches

## Quantitative expression

Transcriptomic signatures

Cancer diagnosis/prognosis/relapse...



[NGuyen et al, 2021]

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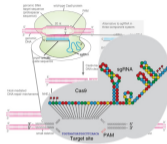
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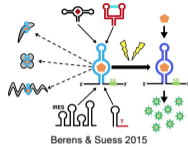
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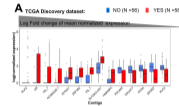
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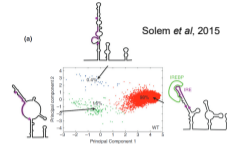
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[NGuyen et al, 2021]



## Non-coding mutations

lncRNAs, miRNAs, structure-associated (RiboSnitches)

$\beta$ -thalassemia, duchenne muscular dystrophy, Cystic fibrosis, Rett syndrome...

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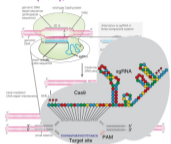


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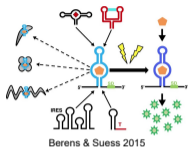
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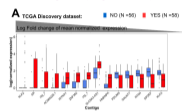
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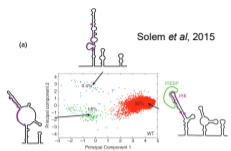
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[NGuyen et al, 2021]



Solem et al, 2015

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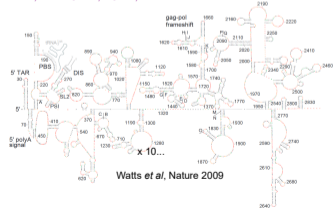
## RiboNucleic Acids (RNAs)



Encodes proteins  
mRNA Vaccines  
COVID-19, Malaria (Zika, CMV, Cancers?)

## Genomic material for Human pathogens

HIV-1, SARS-CoV 2, HCoVs, MERS

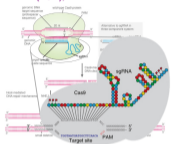


Watts et al, Nature 2009

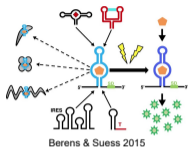
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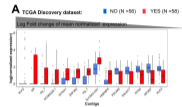
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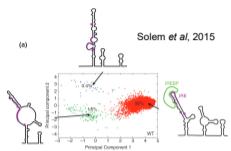
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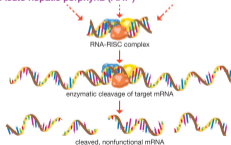
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Cystic fibrosis, Rett syndrome...

## RiboNucleic Acids (RNAs)



## Regulation of gene expression

RNAi therapies (FDA approved)  
Primary hyperoxaluria type 1 (PH1),  
Hereditary transthyretin amyloidosis (ATTRv),  
Acute hepatic porphyria (AHP)



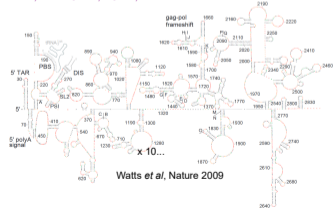
Encyclopaedia Britannica, Inc 2013



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Watts et al, Nature 2009

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Targeting system for DNA Editing

CRISPR therapies

Sickle-cell anemia,  $\beta$ -thalassaemia, Leber congenital amaurosis (LCA), cancer

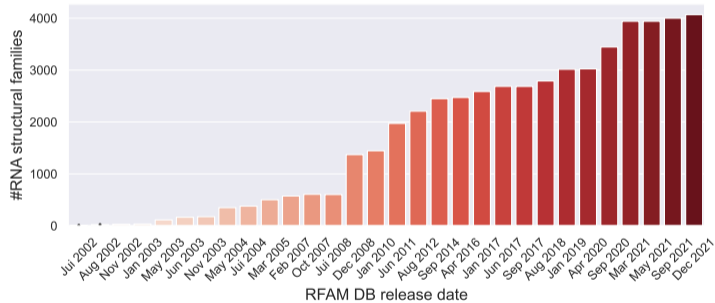


Quantitative expression

Transcriptomic signatures

Cancer diagnosis/prognosis/relapse...

Solem et al. 2015



RNA functional diversity is (largely) enabled by deep structural diversity

Regul  
RNA t  
Primar  
Heredi  
Acute t



devent, 2000

Encyclopaedia Britannica, Inc. 2013

mRNA Vaccines

COVID-19, Malaria (Zika, CMV, Cancers?)

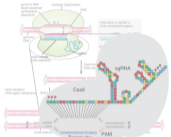
2017

Watts et al. Nature 2009

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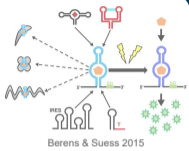
## Rational design

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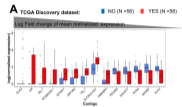


Berens & Suss 2015

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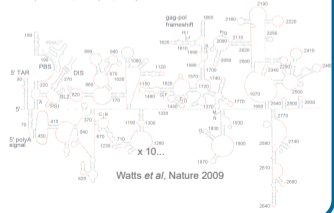
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Cystic fibrosis, Rett syndrome...

## (2D) Structure Modeling

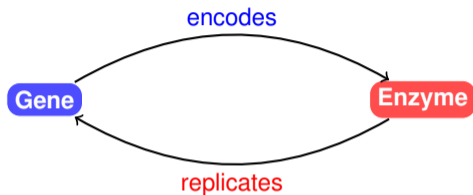
### Genomic material for Human pathogens

HIV-1, SARS-CoV 2, HCoV, MERS



Watts et al, Nature 2009

## RNA world: Resolving the *chicken vs egg* paradox at the origin of life...

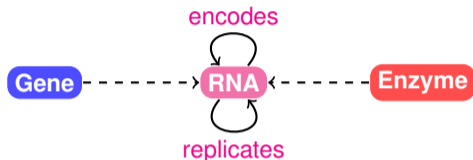


A **gene** big enough to specify **an enzyme** would be too big to replicate accurately without the aid of **an enzyme** of the very kind that it is trying to specify. So the system *apparently cannot get started*.

[...] 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*

## RNA world: Resolving the *chicken vs egg* paradox at the origin of life...



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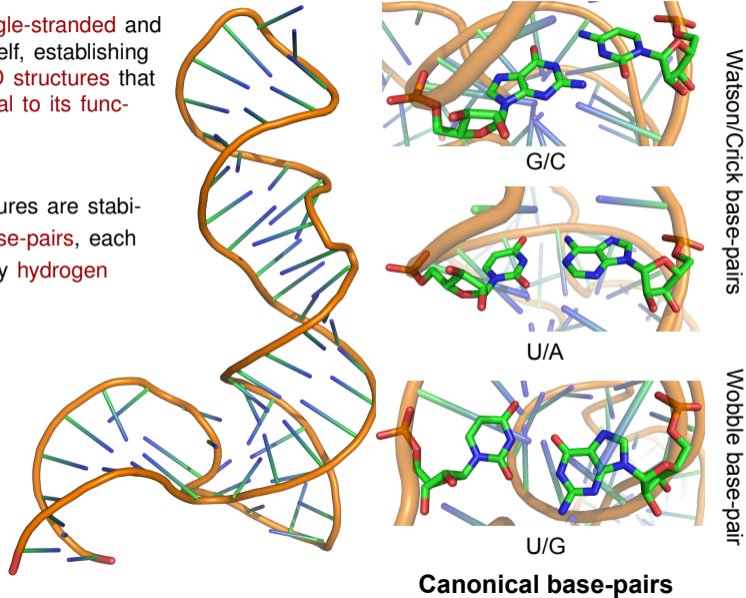
[...] This is the **RNA World**. To see how plausible it is, we need to look at why proteins are good at being enzymes but bad at being replicators; at why DNA is good at replicating but bad at being an enzyme; and finally why *RNA might just be good enough at both roles to break out of the Catch-22*.

**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**.

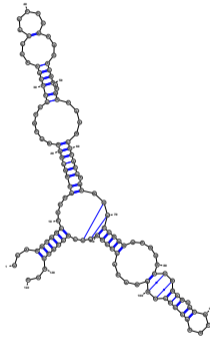


# RNA Structure(s)

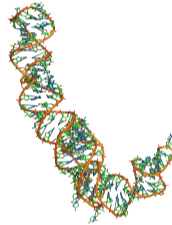
Three<sup>1</sup> levels of representation:

```
UUAGGCGGCCACAGC
GGUGGGGUUGCCUCC
CGUACCCAUCCGAA
CACGGAAGAUAGCC
CACCAGCGUUCGGG
GAGUACUGGAGUGCG
CGAGCCUCUGGAAA
CCCGGUUCGCCCA
CC
```

Primary structure



Secondary structure



Tertiary structure

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

---

<sup>1</sup>Well, mostly...

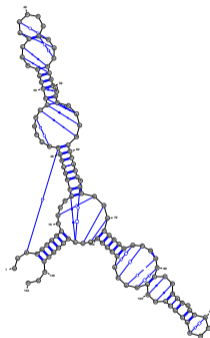


# RNA Structure(s)

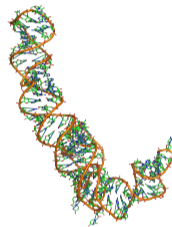
Three<sup>1</sup> levels of representation:

```
UUAGGCGGCCACAGC
GGUGGGGUUGCCUCC
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CGAGCCUCUGGAAA
CCCGGUUCGCCCA
CC
```

Primary structure



Secondary<sup>+</sup> structure



Tertiary structure

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

---

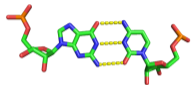
<sup>1</sup>Well, mostly...

## Ignored by secondary structure

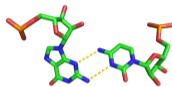
### ► Non-canonical base-pairs

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

**Or** interacting on non-standard edge ( $\neq$  WC/WC-Cis) [LW01].

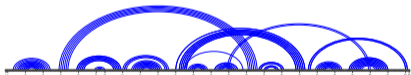


Canonique CG pair(WC/WC-Cis)



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

### ► Pseudoknots (PKs)



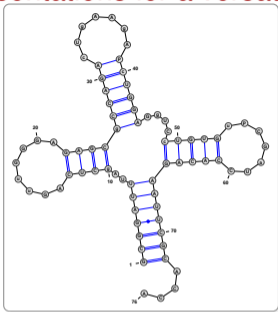
Pseudoknotted structure of group I ribozyme (PDBID: 1Y0Q:A)

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

- Some PK conformations are simply unfeasible;
- Folding *in silico* with general pseudoknots is NP-complete [LP00];

Still, folding on restricted classes of conformations seems promising [CDR<sup>+</sup>04].

# Various representations for a versatile biomolecule



Outer-planar graphs

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

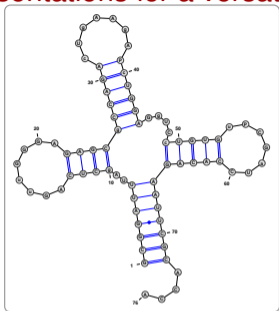
Supporting intuitions

Different representations

Common combinatorial structure

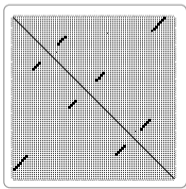
\* Additional steric constraints

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Dot plots

Adjacency matrices\*

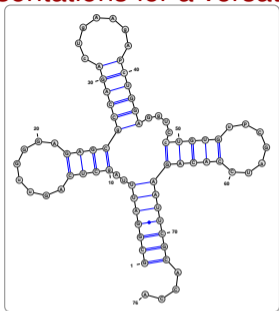
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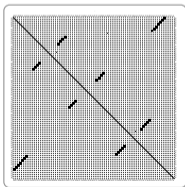
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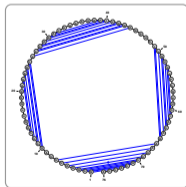


Outer-planar graphs

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



Dot plots  
Adjacency matrices\*



Non-crossing arc diagrams\*

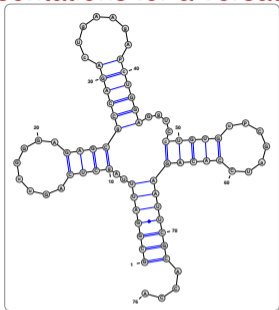
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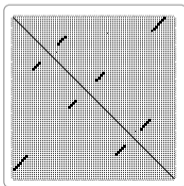


Outer-planar graphs

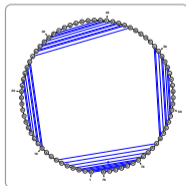
Hamiltonian-path,  $\Delta(G) \leq 3$ , 2-connected\*

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

Motzkin words\*



Dot plots  
Adjacency matrices\*



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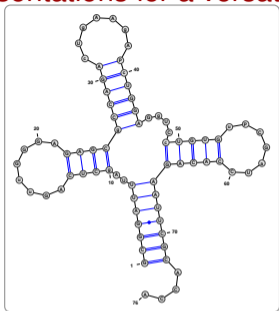
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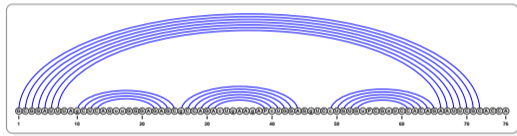


Outer-planar graphs

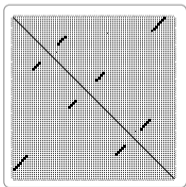
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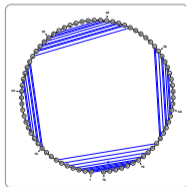
Motzkin words\*



Non-crossing arc-annotated sequences\*



Dot plots  
Adjacency matrices\*



Non-crossing arc diagrams\*

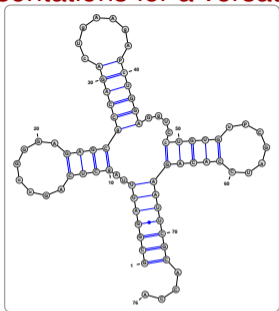
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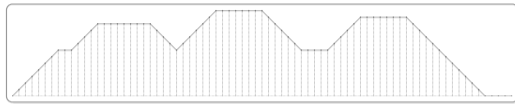


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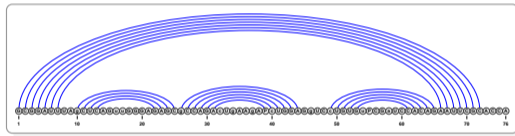
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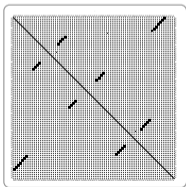
Motzkin words\*



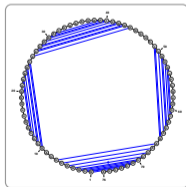
Positive 1D meanders\* over  $S = \{+1, -1, 0\}$



Non-crossing arc-annotated sequences\*



Dot plots  
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# Outline

## Introduction

Dynamic programming 101

Why RNA?

RNA folding

RNA Structure(s)

Some representations of RNA structure

## Some flavours of folding prediction

Thermodynamics vs Kinetics

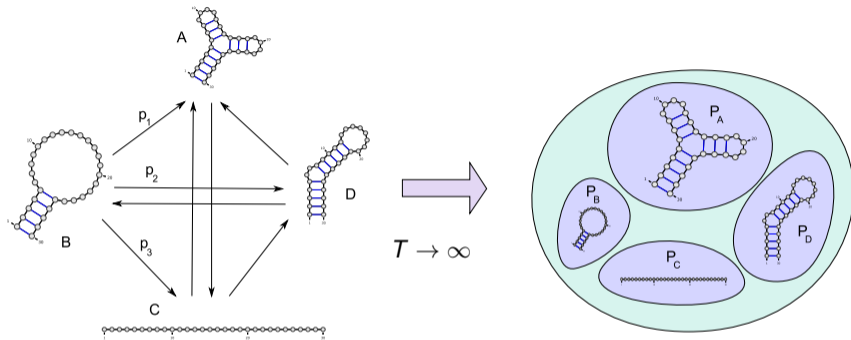
Dynamic programming: Reminder

## Free-energy minimization

Nussinov-style RNA folding

## Thermodynamics *aparté*

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**.

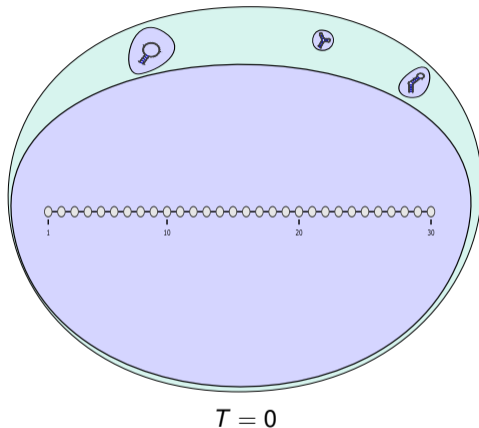
**Corollary:** Initial conformation does not matter.

**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;

## Away from equilibrium

Transcription: RNA synthesized, supposedly without structure<sup>2</sup>



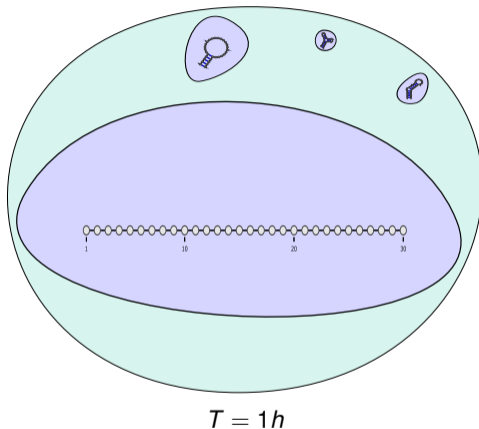
But most mRNAs are degraded before 7h (Org.: Souris [SSN<sup>+</sup>09]).

---

<sup>2</sup>Except for co-transcriptional folding...

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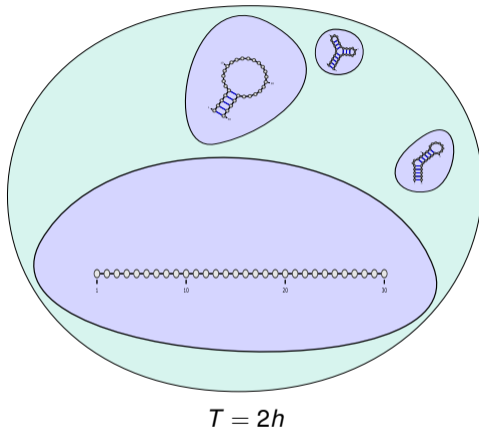
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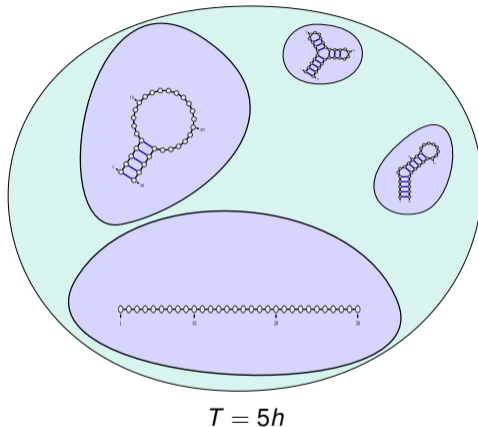
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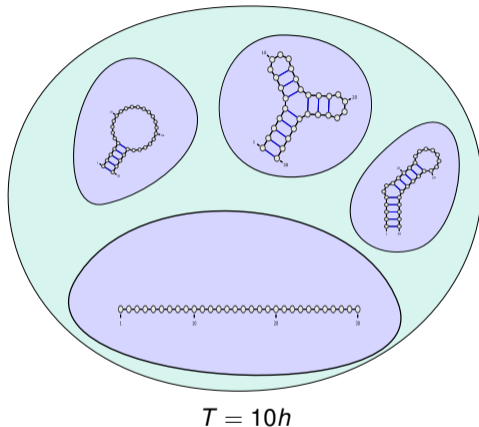
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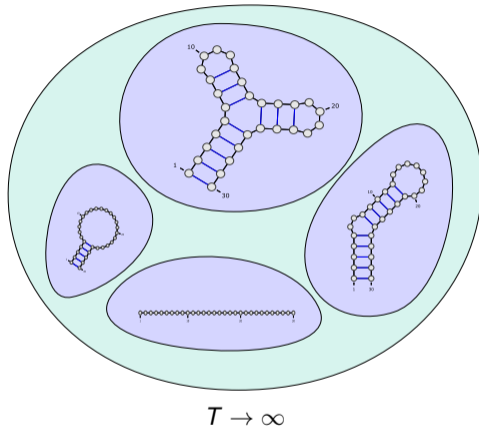
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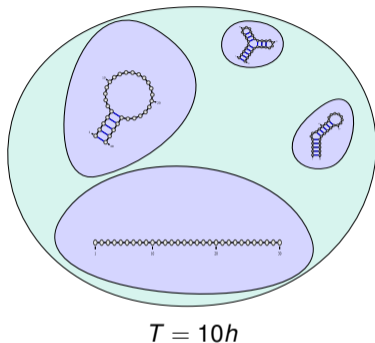
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## Away from equilibrium

Transcription: RNA synthesized, supposedly without structure<sup>2</sup>



But most mRNAs are degraded before 7h (Org.: Souris [SSN<sup>+</sup>09]).

- Determine most stable (Minimum Free-Energy) structure at equilibrium;
- Compute average properties of Boltzmann ensemble;
- Determine most likely structure at finite time  $T$ .

(c.f. H. Isambert through simulation, NP-complete deterministically [MTSC09])

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## Dynamic programming: General principle

Dynamic programming = General optimization technique.

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

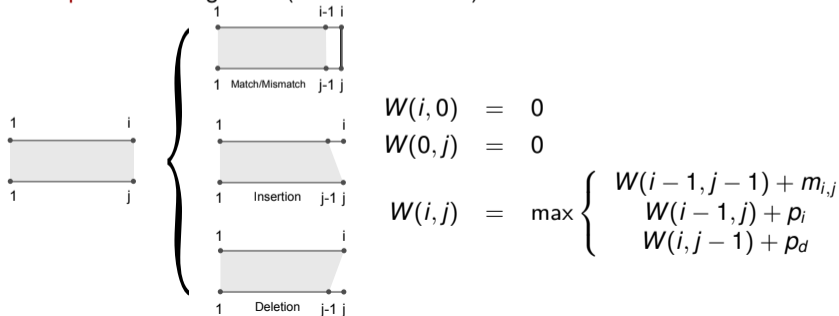
Bioinformatics :

Discrete solution space (alignments, structures...)

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

⇒ Efficient dynamic programming scheme

Example: Local Alignment(Smith/Waterman)



## Algorithmic details

**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								

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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							

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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								

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

Red arrows indicate the path for the local alignment: from (0,0) to (1,2), then to (2,3), and finally to (3,4).

## 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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		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								

Diagram illustrating the dynamic programming table for local alignment. The table shows the alignment of AGCACACA (rows) and ACACACTA (columns). The values in the table represent the maximum score for the alignment up to that point. Red arrows indicate the path of the optimal alignment: A (row 1) aligns with A (col 2), C (row 2) aligns with C (col 3), A (row 3) aligns with A (col 4), and C (row 4) aligns with C (col 5). A grey arrow points from (3,4) to (3,5), and a black arrow points from (3,5) to (4,5).



## Complete example

**Example:** Local alignment of AGCACACA and ACACACTA

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

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

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

		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								

Diagram illustrating the dynamic programming table for local alignment of AGCACACA and ACACACTA. The table shows the maximum score  $W(i, j)$  for each subproblem. Red arrows indicate the path of the optimal alignment: (0,0) to (1,1), (1,1) to (2,2), (2,2) to (3,3), (3,3) to (4,4), (4,4) to (5,5), (5,5) to (6,6), (6,6) to (7,7), (7,7) to (8,8), and (8,8) to (9,9). Grey arrows indicate other possible transitions.

## Complete example

**Example:** Local alignment of AGCACACA and ACACACTA

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

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

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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	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 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								
C	0								
A	0								
C	0								
A	0								

## Complete example

**Example:** Local alignment of AGCACACA and ACACAETA

**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  $p_i = p_j = -1$

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

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

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

		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	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  $p_i = p_j = -1$

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

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

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

		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	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  $p_i = p_j = -1$

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

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

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

Best alignment

A  
A

		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  $p_i = p_j = -1$

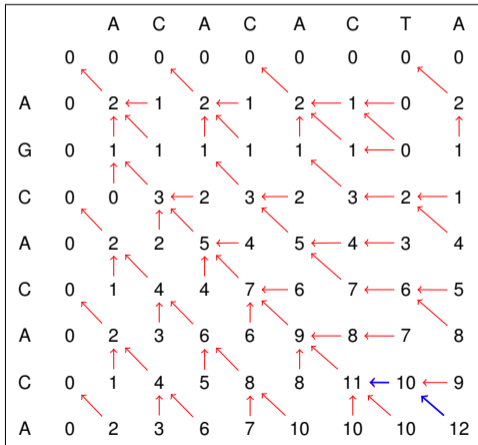
$$W(i, 0) = 0$$

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

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

Best alignment

- A  
T A



# Complete example

**Example:** Local alignment of AGCACACA and ACACACTA

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

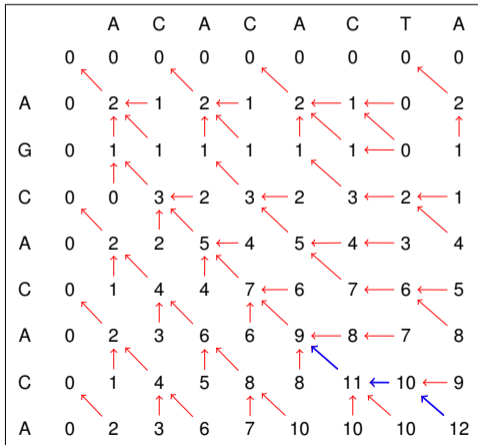
$$W(i, 0) = 0$$

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

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

**Best alignment**

C - A  
C T A



# Complete example

**Example:** Local alignment of AGCACACA and ACACACTA

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

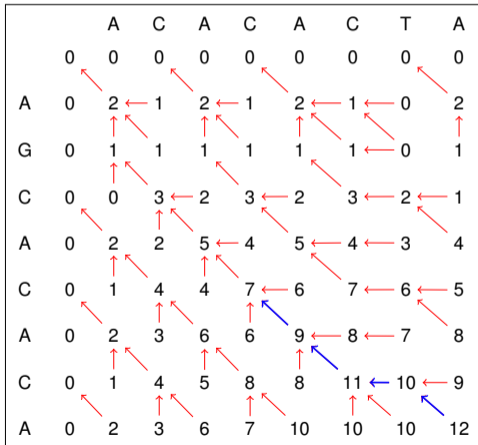
$$W(i, 0) = 0$$

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

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

**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  $p_i = p_j = -1$

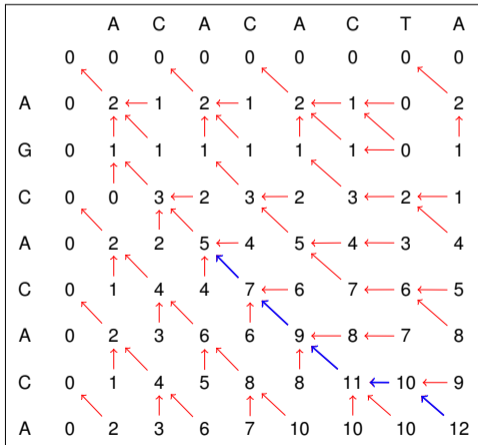
$$W(i, 0) = 0$$

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

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

**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  $p_i = p_j = -1$

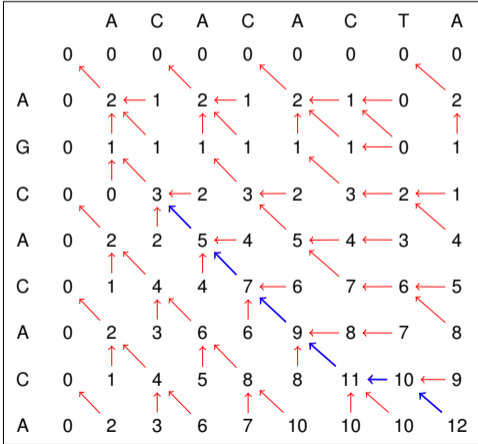
$$W(i, 0) = 0$$

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

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

**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  $p_i = p_j = -1$

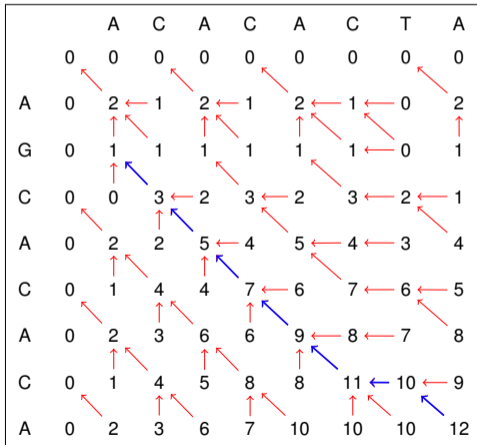
$$W(i, 0) = 0$$

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

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

**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  $p_i = p_j = -1$

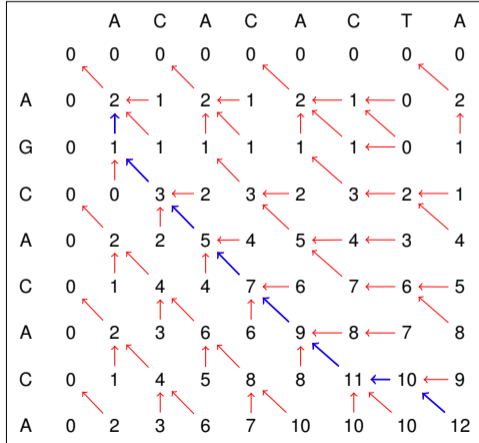
$$W(i, 0) = 0$$

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

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

**Best alignment**

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  $p_i = p_j = -1$

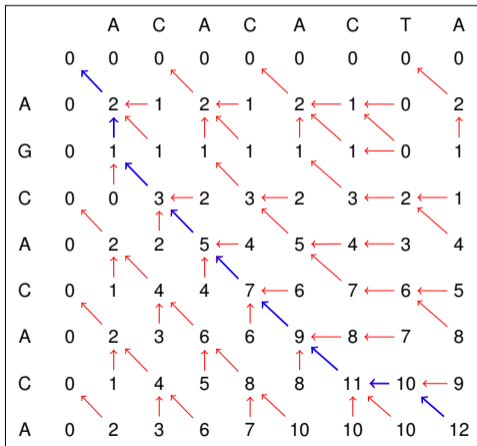
$$W(i, 0) = 0$$

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

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

**Best alignment**

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





# Properties of DP schemes

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.

$\Rightarrow$  Under these properties, one can **enumerate** solution space.

# Outline

## Introduction

Dynamic programming 101

Why RNA?

RNA folding

RNA Structure(s)

Some representations of RNA structure

## Some flavours of folding prediction

Thermodynamics vs Kinetics

Dynamic programming: Reminder

## Free-energy minimization

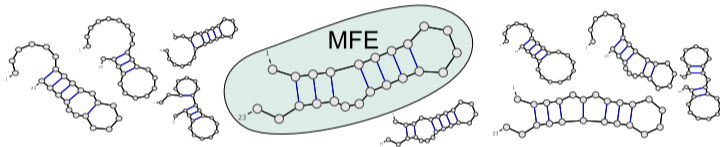
Nussinov-style RNA folding

# Folding by minimizing free-energy

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

Ab initio folding prediction =

Predict RNA structure from its sequence  $\omega$  only.



- ▶ **Conformations:** Set  $S_\omega$  of secondary structures **compatible** (w.r.t. **base-pairing constraints**) with primary structure  $\omega$ .
- ▶ **Free-Energy:** Function  $E_{\omega,S}$  (KCal.mol<sup>-1</sup>), **additive** on motifs occurring in any sequence/conformation couple  $(\omega, S)$ .
- ▶ **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 model

## Nussinov/Jacobson energy model (NJ)

Base-pair maximization (with a twist):

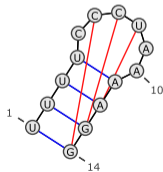
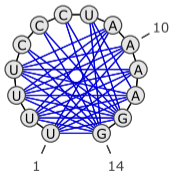
- ▶ Additive model on **independently contributing** base-pairs;
- ▶ **Canonical base-pairs** only: Watson/Crick (A/U,C/G) and Wobble (G/U)

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

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 model

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Base-pair maximization (with a twist):

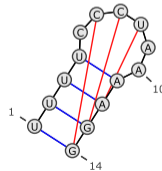
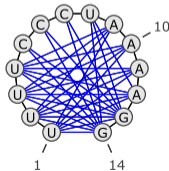
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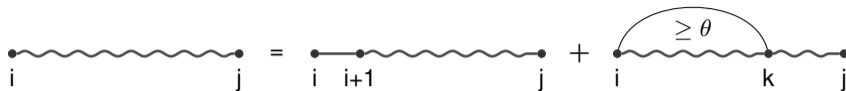
**Variant:** Weight each pair with  $-\#Hydrogen\ bonds$

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

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

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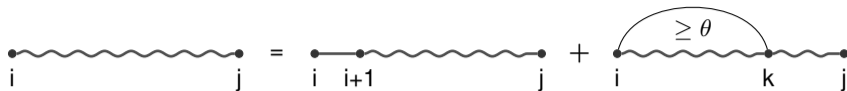
## Nussinov/Jacobson DP scheme



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

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

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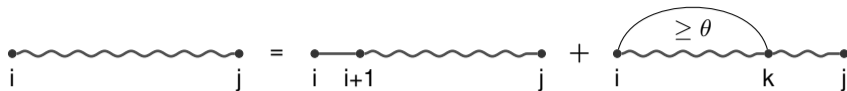
**Correctness.** Goal = Show that MFE over interval  $[i, j]$  is indeed found in  $N_{i,j}$  after completing the computation. Proceed by induction:

- ▶ 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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→ free-energy =  $N_{i+1,j}$

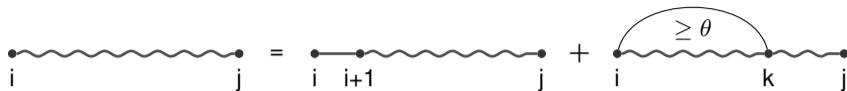
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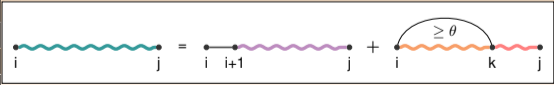
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→ free-energy =  $\Delta G_{i,k} + N_{i+1,k-1} + N_{k+1,j}$

# Nussinov/Jacobson

	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



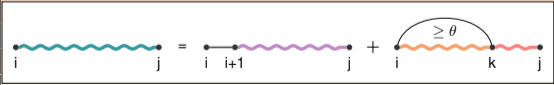
# Nussinov/Jacobson

	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



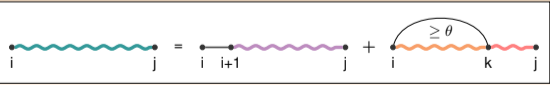
# Nussinov/Jacobson

	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	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	0	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



# Nussinov/Jacobson

	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	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	0	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



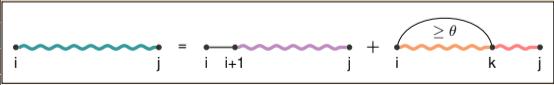
# Nussinov/Jacobson

	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



# Nussinov/Jacobson

	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	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	0	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



# Nussinov/Jacobson

	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





# Nussinov/Jacobson

	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	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	0	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



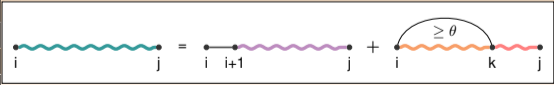
# Nussinov/Jacobson

	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	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	0	3	3	3	5	5
U											0	0	0	0	2	2	2	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



# Nussinov/Jacobson

	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	8	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	0	2	2	2	3	3
U												0	0	0	1	2	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



# Nussinov/Jacobson

	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



# Nussinov/Jacobson

	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



# Nussinov/Jacobson

	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



# Nussinov/Jacobson

	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	



# Nussinov/Jacobson

	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	





# Nussinov/Jacobson

	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	



# Nussinov/Jacobson

	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	



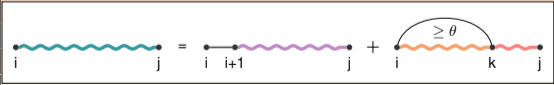
# Nussinov/Jacobson

	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	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	0	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	



# Nussinov/Jacobson

	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	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	0	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	



# Nussinov/Jacobson

	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	
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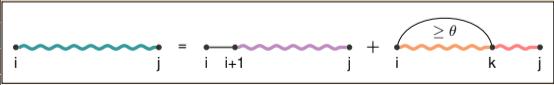
# Nussinov/Jacobson

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



# Nussinov/Jacobson

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



# Nussinov/Jacobson

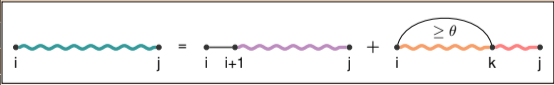
	C	G	G	A	U	A	C	U	U	C	U	U	A	G	A	C	G	A	
	(	(	(	.	.	.	)	.	.	.	.	.	.	.	.	)	)	.	
C	0	0	0	0	0	0	3	4	4	6	6	6	6	9	9	11	14	14	
G		0	0	0	0	0	3	4	4	6	6	6	6	7	9	11	11	11	
G			0	0	0	0	3	3	3	5	5	5	5	6	8	10	10	10	
A				0	0	0	0	2	2	2	2	4	4	5	7	7	8	10	
U					0	0	0	0	0	2	2	4	5	7	7	8	10		
A						0	0	0	0	2	2	2	5	5	5	8	8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





# Nussinov/Jacobson

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



# Nussinov/Jacobson

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



# Nussinov/Jacobson

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



# Nussinov/Jacobson

	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	



# Nussinov/Jacobson

	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	
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