# Complexity and enumerative aspects of multiple RNA design

Stefan Hammer

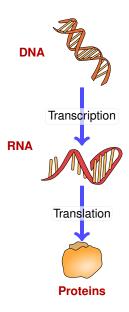
Yann Ponty+,\*

Wei Wang\*,•

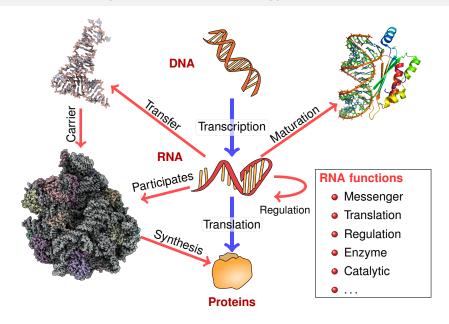
Sebastian Will<sup>†</sup>

- + LIX, CNRS/Ecole Polytechnique
  - \* Amibio team, Inria Saclay
  - LRI, Université Paris-Sud
  - TBI, University of Vienna

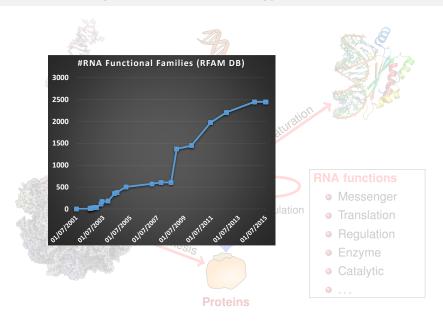
## Fundamental dogma of molecular biology



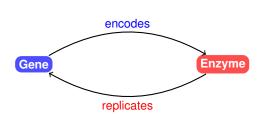
## Fundamental dogma of molecular biology (v2.0)



## Fundamental dogma of molecular biology



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



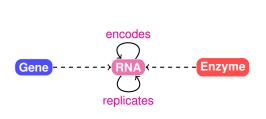


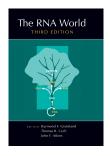
A gene big enough to specify an enzyme would be too big to replicate accurately without the aid of an enzyme of the very kind that it is trying to specify. So the system apparently cannot get started.

[...] This is the RNA World. To see how plausible it is, we need to look at why protein 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...



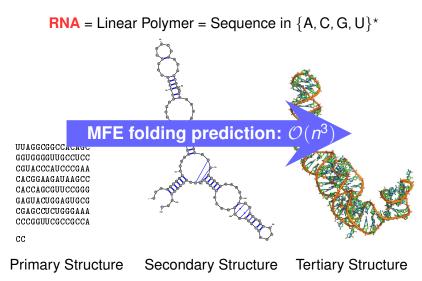


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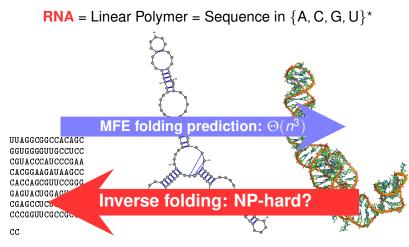
R. Dawkins. The Ancestor's Tale: A Pilgrimage to the Dawn of Evolution

#### RNA sequence and structure(s)



5s rRNA (PDBID: 1K73:B)

#### **RNA** inverse folding



**Primary Structure** 

Secondary Structure

Structure Tertiaire

5s rRNA (PDBID: 1K73:B)

## **Design objectives**

#### Positive structural design

Optimize **affinity** of designed sequences towards target structure Or simply ensure their compatibility with **one or several** structures **Examples:** Most stable sequence for given fold...

#### **Negative structural design**

Limit affinity of designed sequences towards **alternative structures Examples:** Lowest free-energy, High Boltzmann probability/Low entropy...

#### Additional constraints:

- Forbid motif list to appear anywhere in design
- Force motif list to appear each at least once
- Limit available alternatives at certain positions
- Control overall composition (GC-content)

## **Existing approaches for negative design**

#### Based on local search...

- RNAInverse TBI Vienna
- Info-RNA Backofen@Freiburg
- RNA-SSD Condon@UBC
- NUPack Pierce@Caltech
- RNAFBinv Barash@Ben
   Gurion

- ... bio-inspired algorithms...
- ERB Gantjabesh@Tehran
- FRNAKenstein Hein@Oxford
- AntaRNA Backofen@Freiburg
- ... exact approaches...
- RNAIFold Clote@Boston College
- CO4 Will@Vienna

RNA negative design remains a very active area of research ...

... whose computational complexity remains largely unknown!

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**Examples:** Most stable sequence for given fold...

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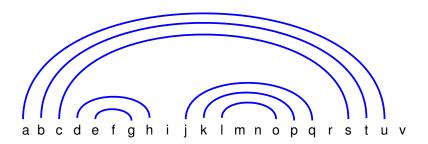
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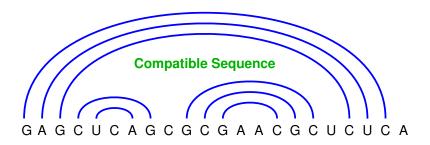


Compatible Base Pairs = Only Watson-Crick base pairs



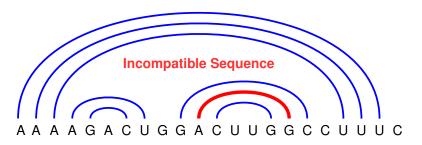


Compatible Base Pairs = Only Watson-Crick base pairs





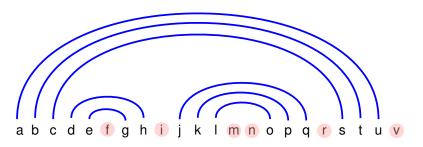
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**Question:** How many **Compatible** sequences?



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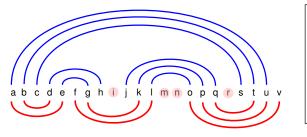


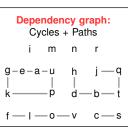
**Question:** How many **Compatible** sequences?

**Answer:**  $4^{\text{\#BPs}} \times 4^{\text{\#Unpaired}} \rightarrow 268435456$ 



Compatible Base Pairs = Only Watson-Crick base pairs

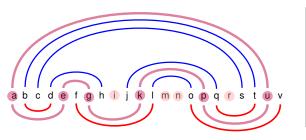


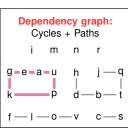


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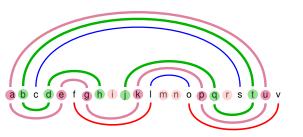


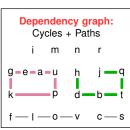


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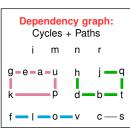


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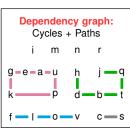


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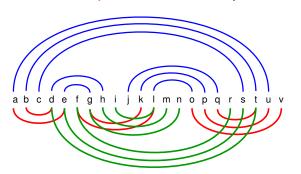


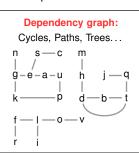
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$$4^{\#CCs} \rightarrow 65536$$



#### Compatible Base Pairs = Only Watson-Crick base pairs

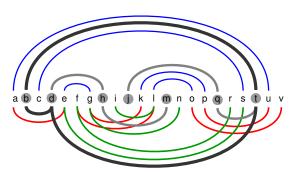


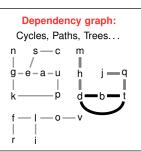


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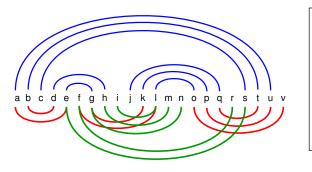


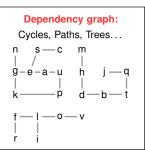


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Compatible Base Pairs = Only Watson-Crick base pairs

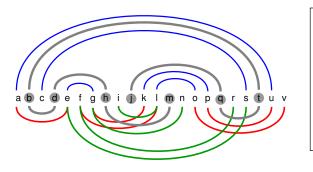


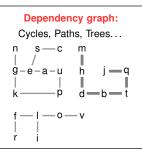


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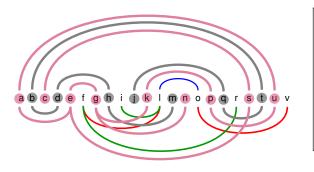


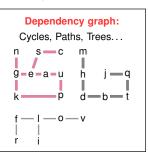


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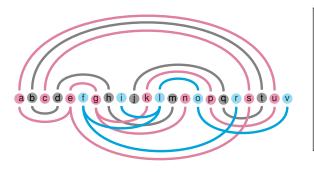


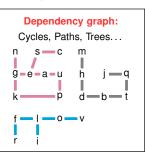


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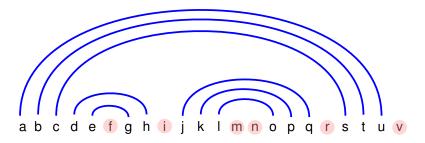




**Question:** How many **Compatible** sequences?



Compatible Base Pairs = Include Wobble base pairs



**Question:** How many **Compatible** sequences?

**Answer:** 4<sup>#Unpaired</sup> × 6<sup>#BPs</sup> → 6 879 707 136



Compatible Base Pairs = Include Wobble base pairs

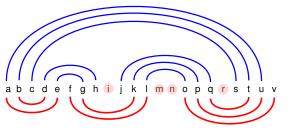


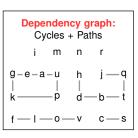
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**Answer:**  $4^{\text{#Unpaired}} \times 6^{\text{#BPs}} \rightarrow 6879707136$ 



Compatible Base Pairs = Include Wobble base pairs

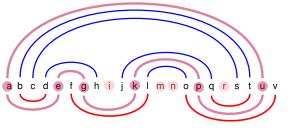


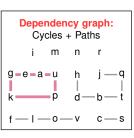


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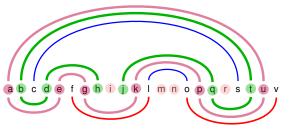


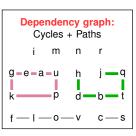


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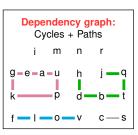


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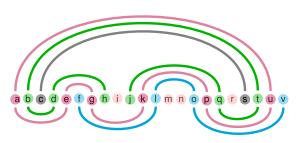


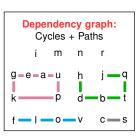


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$$\#\mathsf{Designs}(\mathit{G}) = \prod_{\mathit{c} \in \mathit{CC}(\mathit{G})} \#\mathsf{Designs}(\mathit{cc})$$

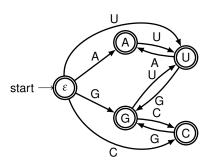
#### Theorem (#Compatible designs for paths and cycles)

The numbers p(n) and c(n) of compatible designs for paths and cycles of length n are:

$$p(n) = 2 \mathcal{F}_{n+2}$$
 and  $c(n) = 2 \mathcal{F}_n + 4 \mathcal{F}_{n-1}$ 

where  $\mathcal{F}_n$  is the n-th Fibonacci number, s.t.  $\mathcal{F}_0 = 0$ ,  $\mathcal{F}_1 = 1$  and  $\mathcal{F}_n = \mathcal{F}_{n-1} + \mathcal{F}_{n-2}$ .

For paths: A simple DFA generates compatible sequences



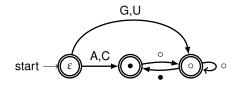
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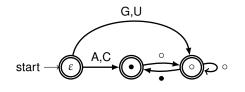
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For paths: A simple DFA generates compatible sequences



$$m_{\bullet}(n) = m_{\circ}(n-1)$$

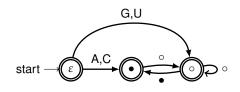
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$$\begin{split} m_{\bullet}(n) &= m_{\circ}(n-1) \\ m_{\circ}(n) &= m_{\circ}(n-1) + m_{\bullet}(n-1) \\ &= m_{\circ}(n-1) + m_{\circ}(n-2) \\ &= \mathcal{F}(n+2) \end{split}$$

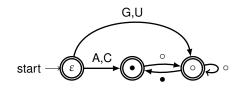
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For paths: A simple DFA generates compatible sequences



$$m_{\bullet}(n) = m_{\circ}(n-1)$$
  
 $m_{\circ}(n) = m_{\circ}(n-1) + m_{\bullet}(n-1)$   
 $= m_{\circ}(n-1) + m_{\circ}(n-2)$   
 $= \mathcal{F}(n+2)$ 

(Since 
$$m_{\circ}(0) = 1$$
 and  $m_{\circ}(1) = 2$ )

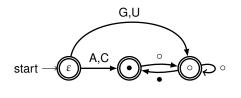
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(Since 
$$m_{\circ}(0) = 1$$
 and  $m_{\circ}(1) = 2$ )

$$p(n) := m_{\varepsilon}(n) = 2 m_{\bullet}(n-1) + 2 m_{\circ}(n-1) = 2(\mathcal{F}(n) + \mathcal{F}(n+1)) = \mathcal{F}(n+2)$$

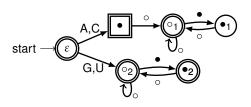
#### Theorem (#Valid designs for paths and cycles)

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For cycle: A barely more involved DFA generates compatible sequences



$$m_{\circ_2}(n) = \mathcal{F}(n+2)$$
  
 $m_{\circ_1}(n) = \mathcal{F}(n+1)$ 

(Since 
$$m_{\circ_1}(0) = 1$$
 and  $m_{\circ_1}(1) = 1$ )

$$c(n) := m_{\varepsilon}(n) = 2 m_{o_1}(n-2) + 2 m_{o_2}(n-1)$$
  
=  $2(\mathcal{F}(n-1) + \mathcal{F}(n+1)) = 2 \mathcal{F}(n) + 4 \mathcal{F}(n-1)$ 

#### Theorem (#Valid designs for paths and cycles)

The numbers p(n) and c(n) of compatible designs for paths and cycles of length n are:

$$p(n) = 2 \mathcal{F}_{n+2}$$
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where  $\mathcal{F}_n$  is the n-th Fibonacci number, s.t.  $\mathcal{F}_0=0$ ,  $\mathcal{F}_1=1$  and  $\mathcal{F}_n=\mathcal{F}_{n-1}+\mathcal{F}_{n-2}$ .

### Theorem (#Compatible designs for general 2-structures graphs)

Let G be the dependency graph associated with 2 RNA structures (max degree=2). The number #Designs(G) of compatible designs for G is given by

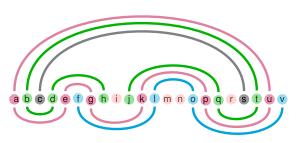
$$\#\mathsf{Designs}(\mathit{G}) = \prod_{\mathit{p} \in \mathcal{P}(\mathit{G})} 2\,\mathcal{F}_{|\mathit{p}|+2} \times \prod_{\mathit{c} \in \mathcal{C}(\mathit{G})} \left(2\,\mathcal{F}_{|\mathit{c}|} + 4\,\mathcal{F}_{|\mathit{c}|-1}\right)$$

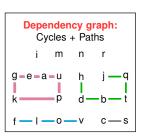
where G decomposes into paths  $\mathcal{P}(G)$  and cycles  $\mathcal{C}(G)$ .

# Counting compatible sequences: WC/Wobble + Two structures



Compatible Base Pairs = Include Wobble base pairs





**Question:** How many **Compatible** sequences?

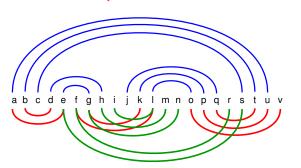
**Answer:**  $\neq \varnothing$ ! (base-pairs and dependency graphs always bipartite)

$$\# \mathsf{Designs}(\mathit{G}) = \prod_{c \in \mathit{CC}(\mathit{G})} \# \mathsf{Designs}(\mathit{cc}) = 2322432$$

# **Counting compatible sequences: Watson-Crick +** > 2 structures



#### Compatible Base Pairs = Include Wobble base pairs



# Dependency graph: Cycles, Paths, Trees... n s—c m g-e-a-u h j—q | | | | | k——P d—b—t f—I—o—v r i

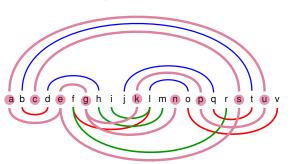
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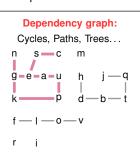
**Answer:** Non-bipartite  $\rightarrow \varnothing$ ; Bipartite  $\rightarrow$ 

# **Counting compatible sequences: Watson-Crick +** > 2 structures



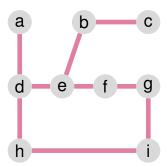
#### Compatible Base Pairs = Include Wobble base pairs



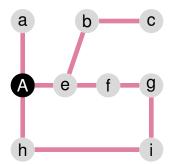


**Question:** How many **Compatible** sequences?

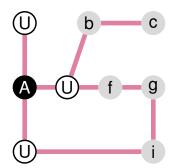
Answer: Non-bipartite 
$$\to \varnothing$$
; Bipartite  $\to \prod_{cc \in CC(G)} 2 \times \#IS(cc)$ 



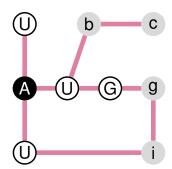




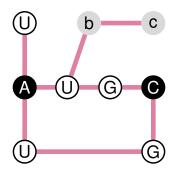




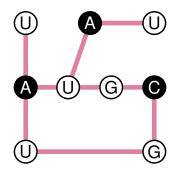




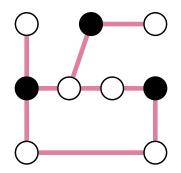


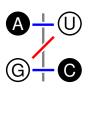






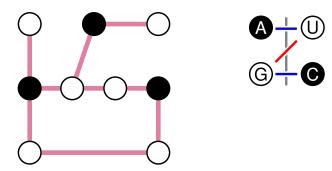






**Remark:** No adjacent **black letters** in compatible designs Up to trivial symmetry\* (e.g. top-left position  $\in \{G, A\}$ ):

Designs\* $(cc) \subseteq IndependentSets(cc)$ 

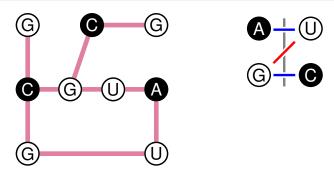


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Also, IS (black vert.) +  $\nwarrow$  vert.  $\in$  {G, A}  $\Rightarrow$  Unique compatible design



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 $\Rightarrow$  Bijection between Designs\*(cc) and IndependentSets(cc).

#### Theorem (#Valid design for bipartite connected dependency graphs)

Let G be a bipartite connected dependency graph, one has:

$$\# \mathsf{Designs}(\mathsf{G}) = 2 \times \# \mathsf{Designs}^{\star}(\mathsf{G}) = 2 \times \# \mathsf{IS}(\mathsf{G})$$

For a bipartite dependency graph G is then

$$\#Designs(G) = \prod_{cc \in CC(G)} 2 \times \#IS(cc) = 2^{|CC(G)|} \times \#IS(G)$$

But #IS(G) is #P-hard on bipartite graphs [Bubbley&Dyer'01] (+ Any G is a dependency graph)

Algorithm  $A \in P$  for  $\#Designs(G) \to Algorithm <math>A' \in P$  for #BIS.

Theorem

#Designs is #P-haro

No polynomial algorithm for #Designs(G) unless  $\#P = FP \ (\Rightarrow P = NP)$ 

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#### **Theorem**

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No polynomial algorithm for #Designs(G) unless  $\#P = FP \ (\Rightarrow P = NP)$ 

#### Consequences

#### **Corollary (#Approximability for** ≤ 5 **structures)** [Weitz'06]

For any G built from  $\leq$  5 pseudoknotted structures, #Design(G) can be approximated within any ratio in polynomial time (PTAS)

#### Corollary (#BIS hardness for > 5 struct.) [Cai, Galanis, Goldberg, Jerrum, McQuillan'16]

Beyond 5 **pseudoknotted** structures, approximating #Design becomes **as hard as** approximating #BIS without any constraint.

Why pseudoknotted? Because any bipartite graph of max degree  $\Delta$  can be decomposed into  $\Delta$  matchings in polynomial time (Vizing's theorem).

Finally, strong connection between **counting** and **sampling** [Jerrum, Valiant, Vazirani'86]

#### Conjecture (#BIS hardness of sampling)

Generating compatible sequences (almost) uniformly w.r.t. a set of structures is #BIS-hard.

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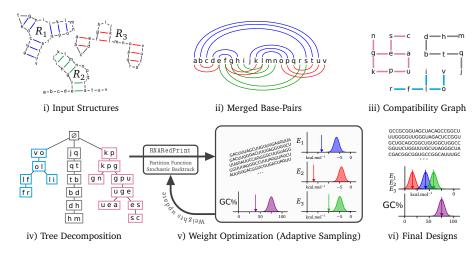
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# Perspectives: FPT and Boltzmann sampling algorithms

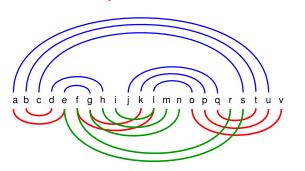


- FPT algorithm for counting based on tree decomposition
- Multidimensional Boltzmann sampling to control energies, GC...

# **Counting compatible sequences: Watson-Crick +** > 2 structures



#### Compatible Base Pairs = Include Wobble base pairs



# 

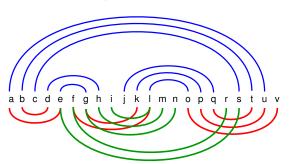
**Question:** How many **Compatible** sequences?

**Answer:** Bipartite →

# **Counting compatible sequences: Watson-Crick +** > 2 structures



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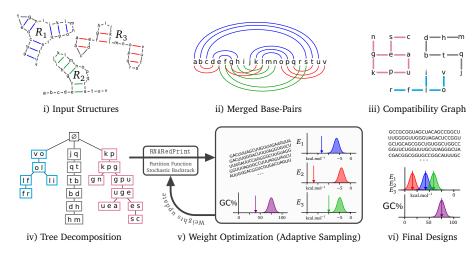


# 

**Question:** How many **Compatible** sequences?

Answer: Bipartite 
$$\rightarrow \prod_{cc \in CC(G)} 2 \times \#IS(cc) = 496\,672$$

# Perspectives: FPT and Boltzmann sampling algorithms



- FPT algorithm for counting based on tree decomposition
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# Thanks!



Submission deadline Nov 6<sup>th</sup> Registration open soon...