

# RNA design in theory and practice

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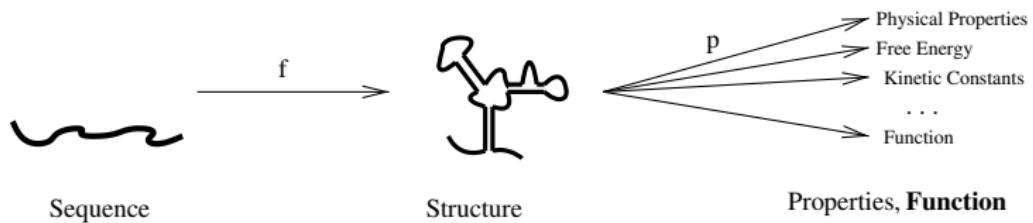
Bioinformatics Group and Interdisciplinary Center for Bioinformatics,  
Department of Computer Science,  
University Leipzig

AlgoSB 2019

January, 2019

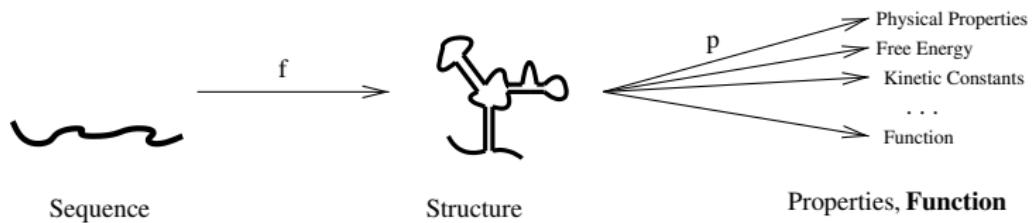
# Sequence-Structure-Function Relation

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Structure of an RNA sequence typically determines the function.**



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# Structure-Puzzle

See a structure...

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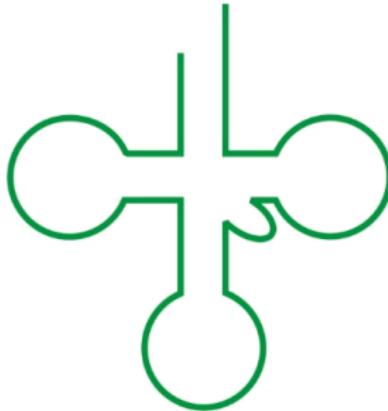
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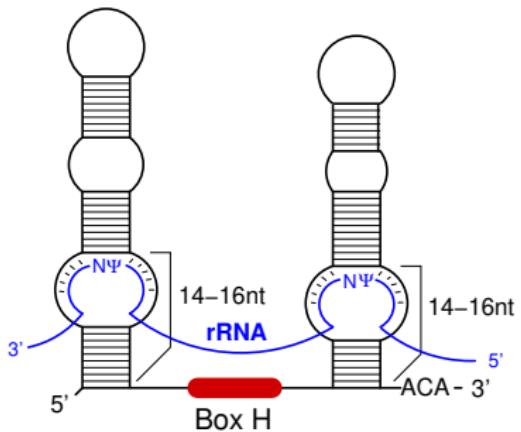


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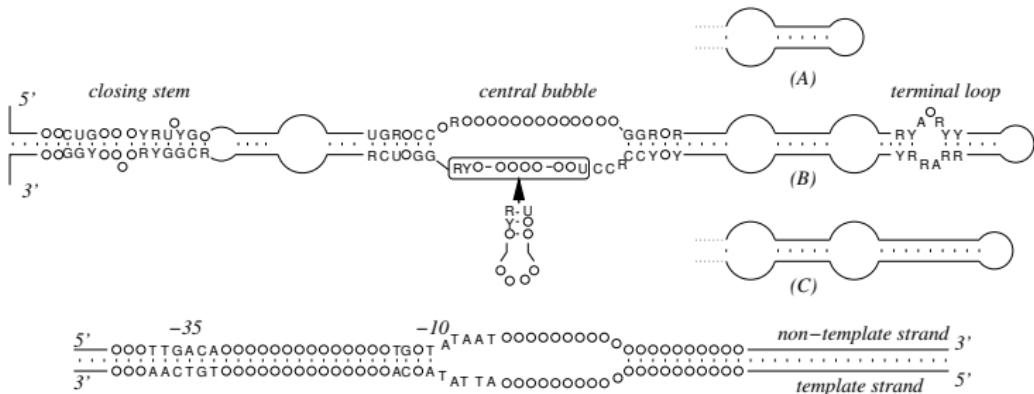


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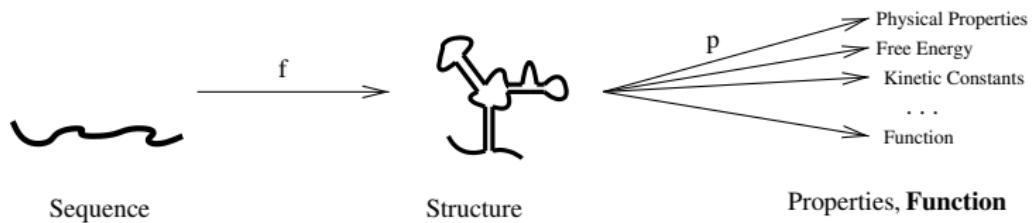


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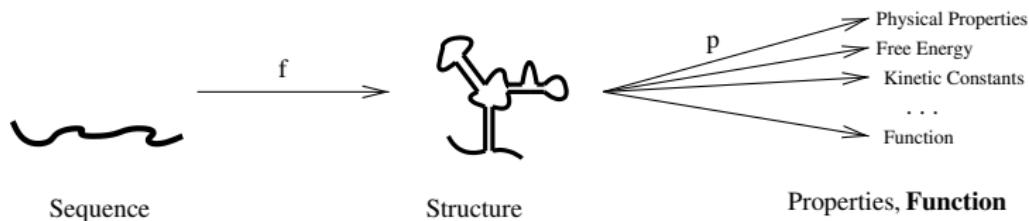
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- ① Which sequence optimally folds into given target structure(s)?
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# Complexity of Sequence Design I

- **Nucleic acid sequence sampling:** four possible assignments on each position.

NNNNNNNNNNNN =  $4^{12} = 16,777,216$  possible sequences

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- **Combination of sequence and structure constraint:**

$$(((((\dots)))).)$$

$$\text{NGNNNNVNNNN} = 3 \times 4^3 \times 2 \times 6^3 = 82,944 \text{ possible sequences}$$

# Sequence Selection

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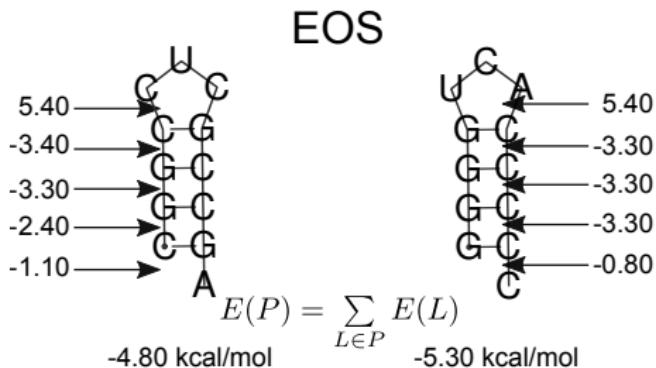
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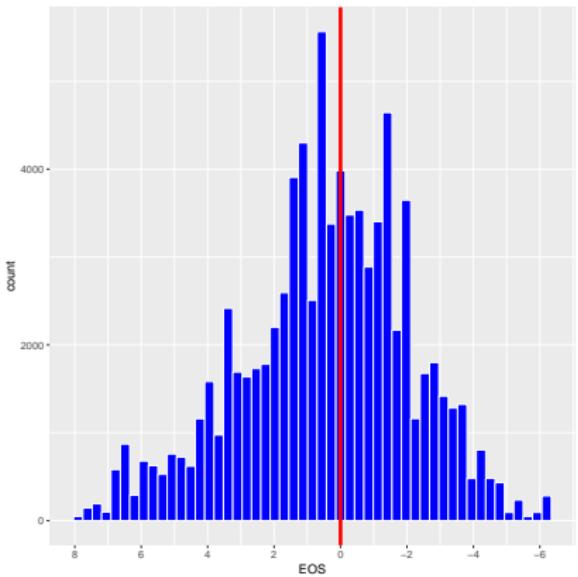
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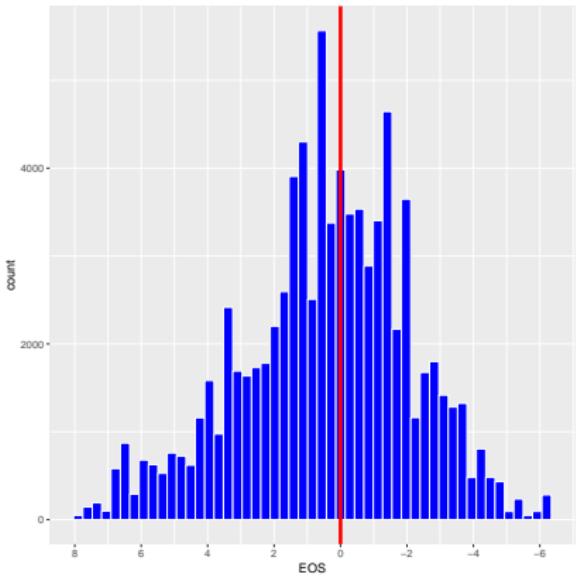
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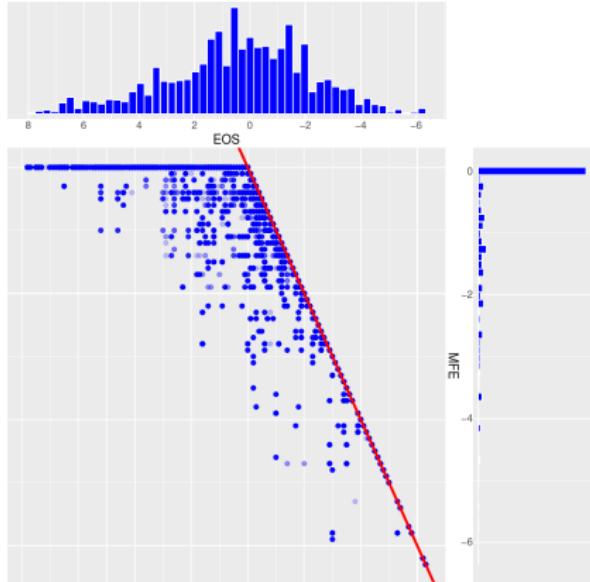
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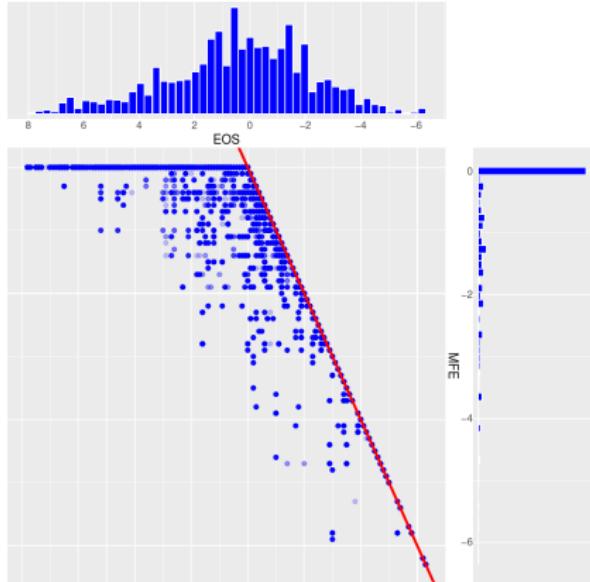
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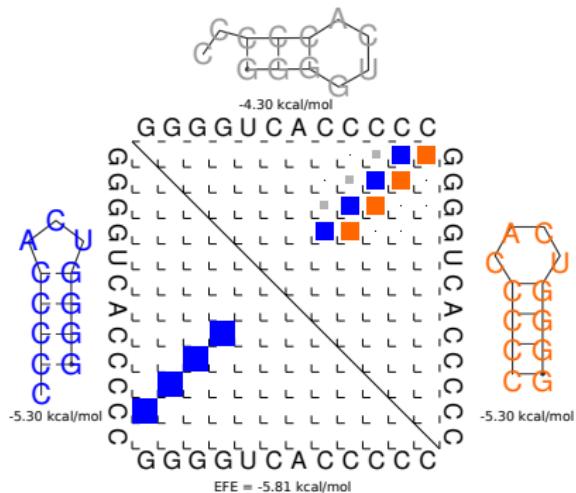
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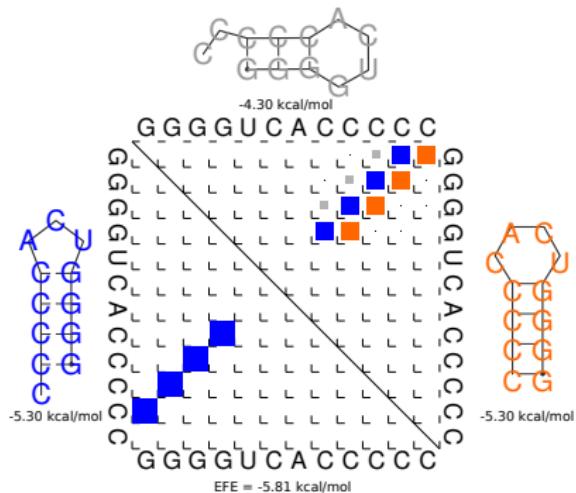
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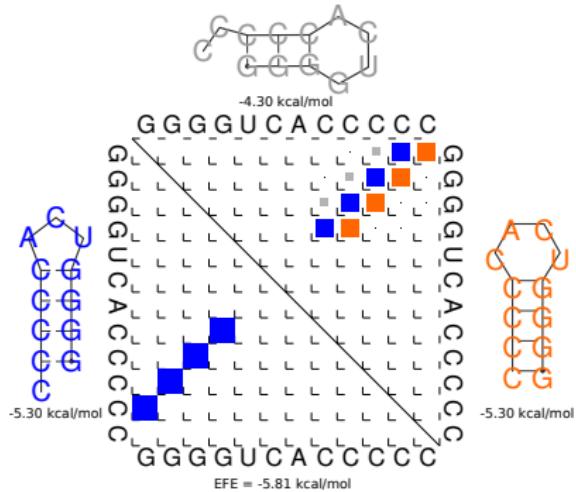
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Does the target conformation dominate the ensemble?

$$EFE = -RT \ln \underbrace{\sum_P \exp\left(-\frac{E(P)}{RT}\right)}_{\text{partition function } Z}$$

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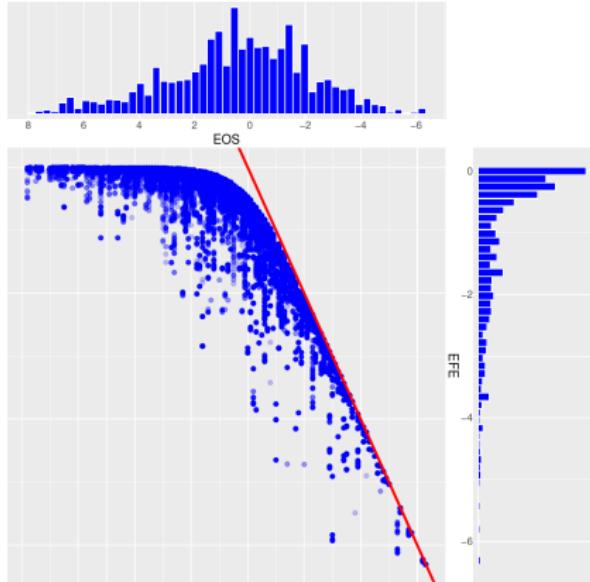
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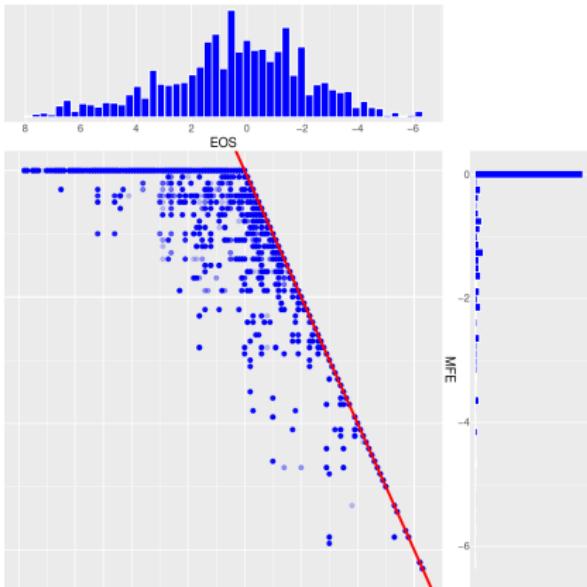
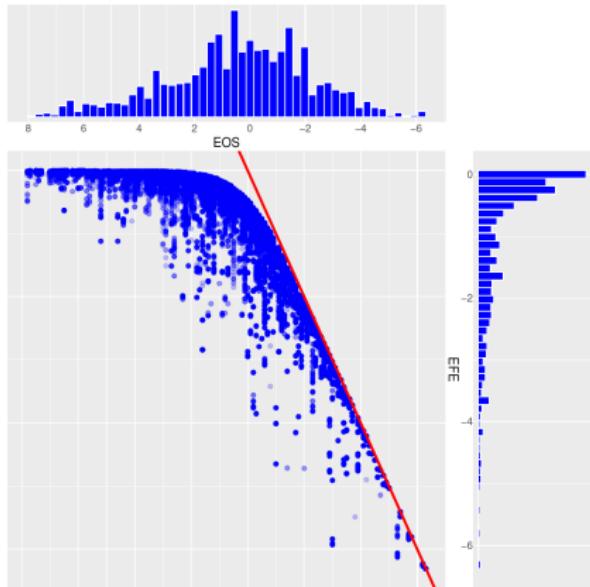
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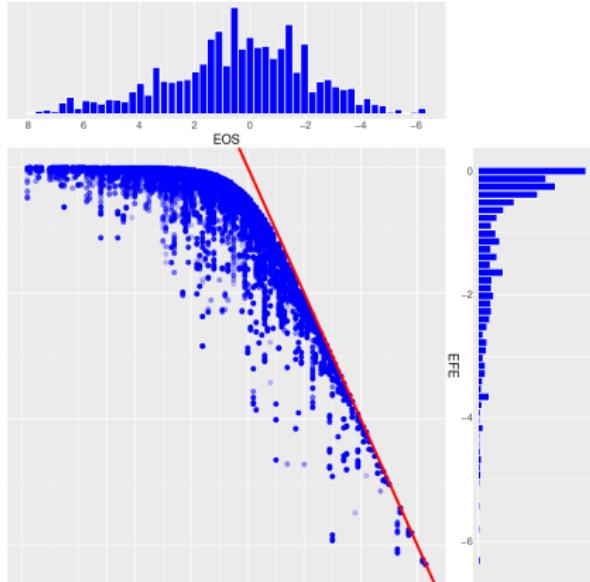
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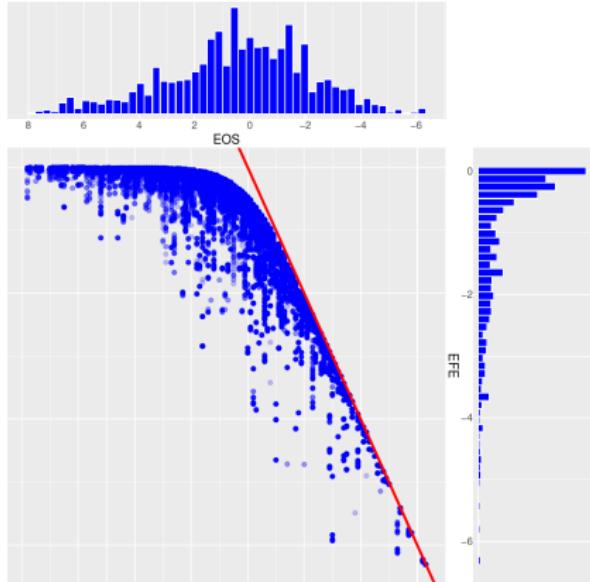
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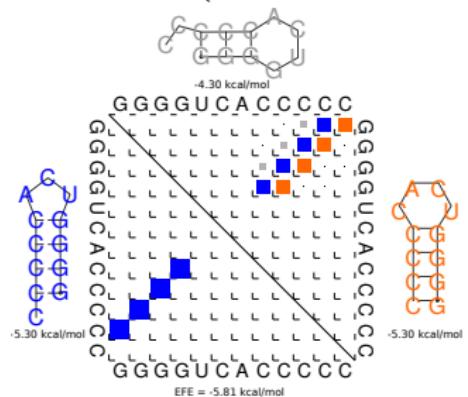
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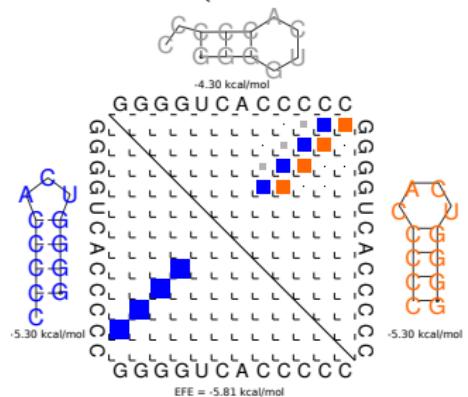
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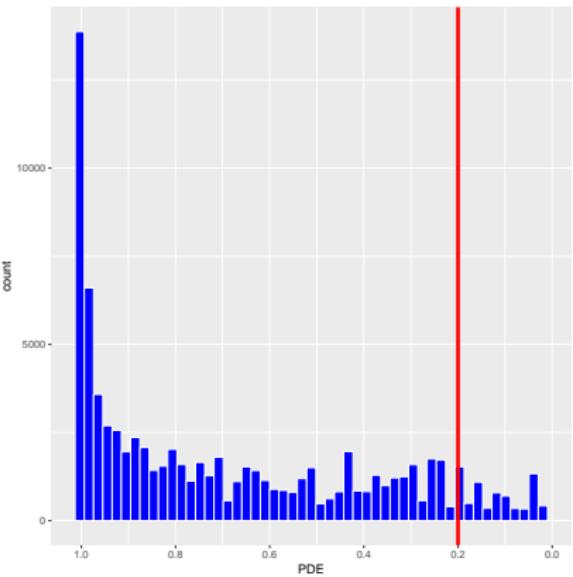
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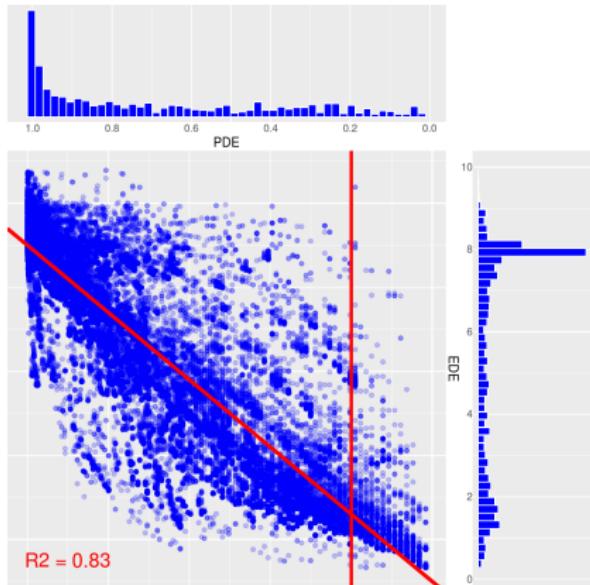
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- favor low EOS
- favor low EOS = MFE
- MFE state rarely the only stable conformation
- structure ensemble might be diverse
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- PDE takes only the exact target state into account
- EDE also counts similar structures

Zadeh, Joseph N., et al., 2011, Sequence Design via Efficient Ensemble Defect Optimization **J Comput Chem** 32: 439–452

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- 1994 RNAinverse
- 2004 RNA-SSD
- 2006 INFO-RNA
- 2011 RNAexinv
- 2011 NUPACK:Design\*
- 2012 RNA-ensign
- 2012 DSS-Opt
- 2013 RNAiFold\*
- 2013 IncaRNAtion
- 2014 ERD
- 2014 EteRNA(Bot)\*
- 2015 antaRNA
- ...

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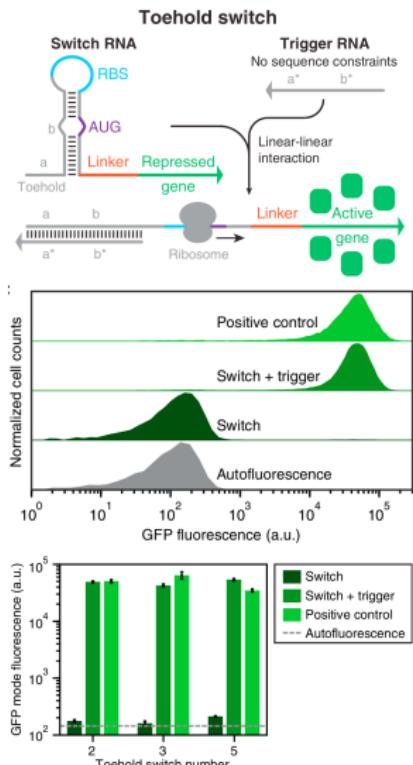
# **“Nature scores!”**

Rhiju Das

# Ensemble Defect based Design

sRNA binding regulates **accessibility** of region around the RBS and AUG and therefor **activates expression**

- **NUPACK:Design** for optimization towards **ensemble defect**
- **GFP reporter assays** with flow cytometry
- **Iterative design approach:** Tested various loop sizes, trigger RNA lengths, binding site length and their effect
- **in silico screening** to predict cross-talk, free energy of structural features, duplex formation, stability of toehold switch region



Green, A., et al., 2014, Toehold Switches: De-Novo-Designed Regulators of Gene Expression, **Cell**, 159, 925–939

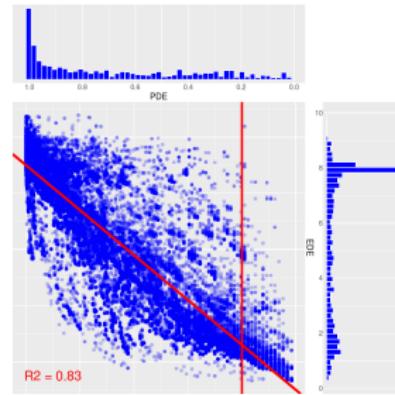
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- Combination of sequence and **one** structure constraint

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$\text{NGNNNNVNNNN} = 3 \times 4^3 \times 2 \times 6^3 = 82,944$   
possible sequences

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$\text{GRRRNNVYYYYY} = 1 \times 2^8 \times 4^2 \times 3 = 12,288$   
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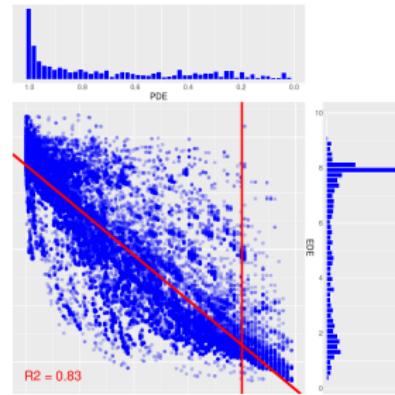
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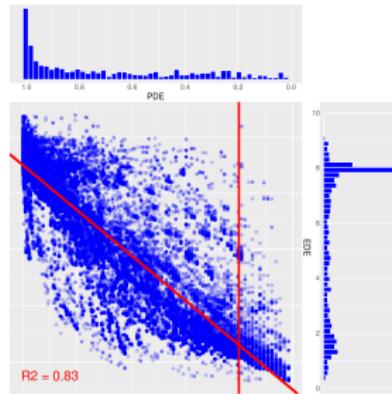
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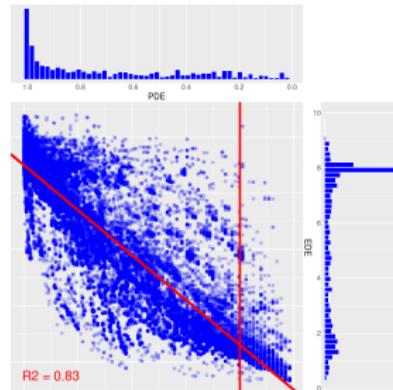
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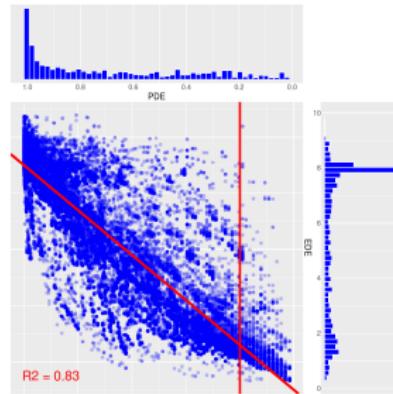
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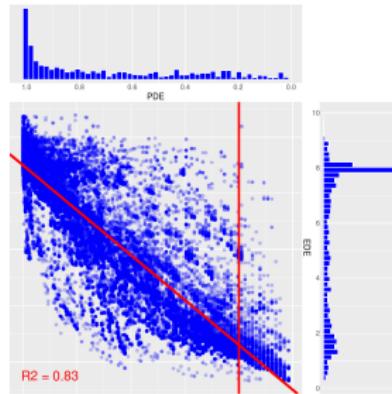
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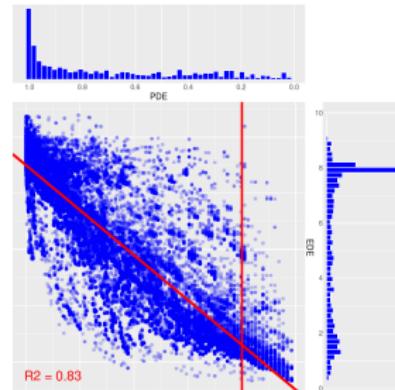
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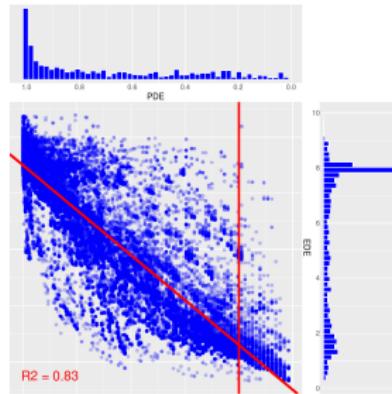
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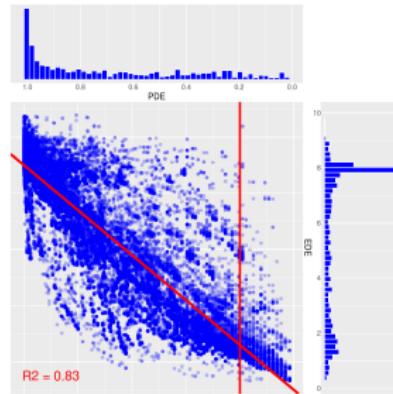
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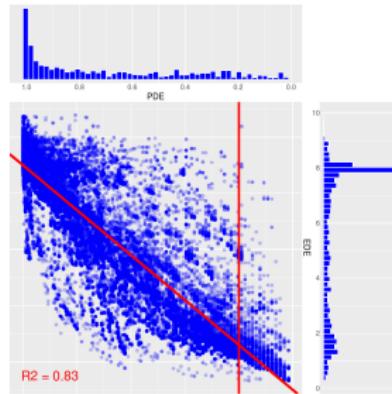
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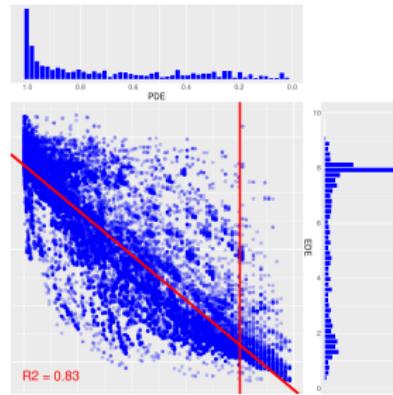
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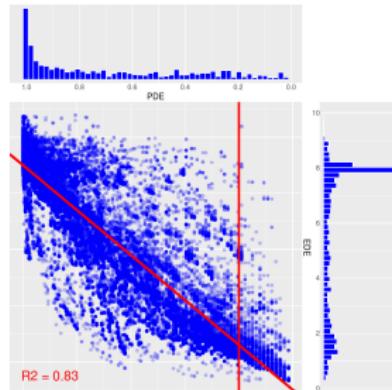
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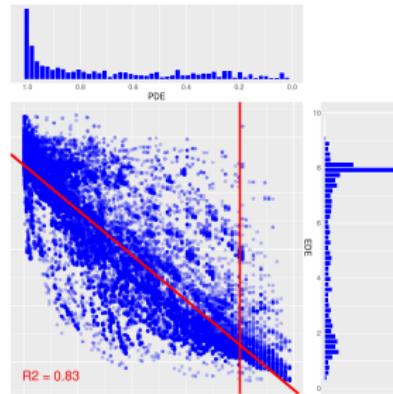
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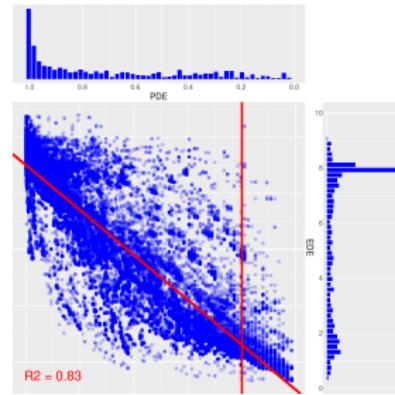
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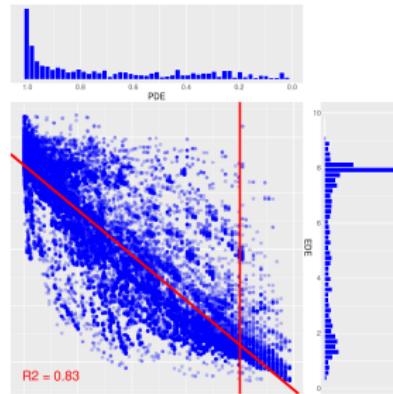
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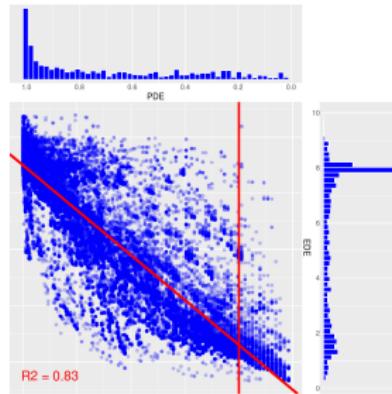
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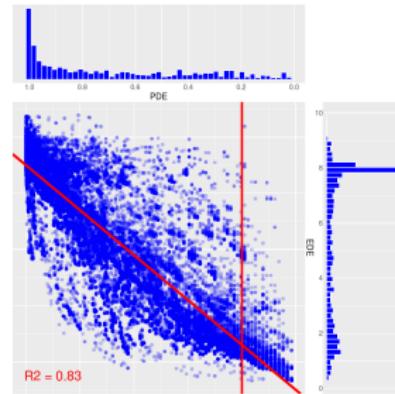
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NNNNNNNNNNNN

1.....6.....12

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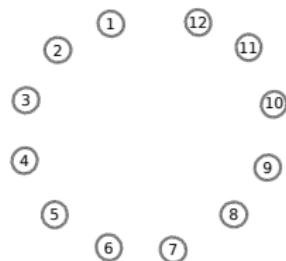
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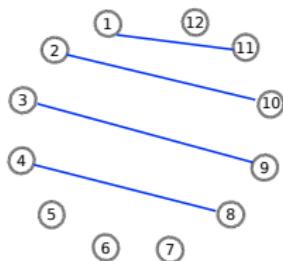
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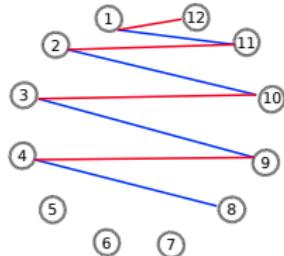
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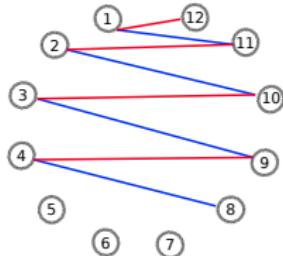
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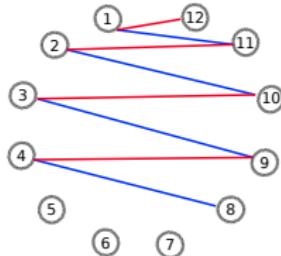
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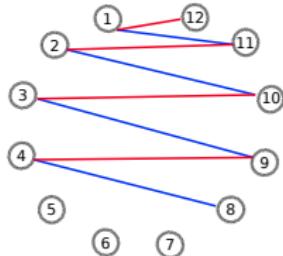
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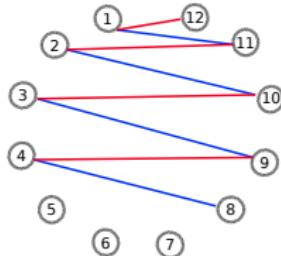
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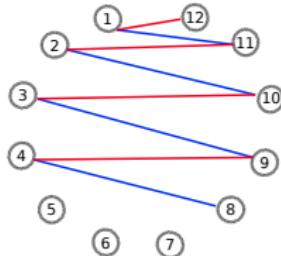
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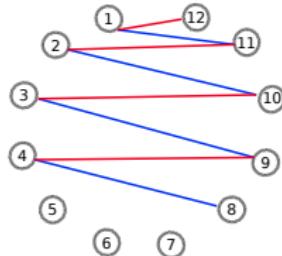
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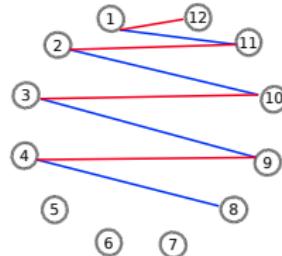
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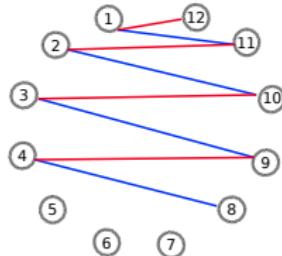
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NNNNNNNNNNNN

1 . . . . 6 . . . . 12



- Possible assignments for a base pair:

$$\sum_{i,j} BPT[i, j] = 6$$

- Possible assignments for a path of length two:

- Possible assignment for a path of length  $L$ :

$$BPT = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \end{bmatrix}$$

( )

A U

C G

G C

G U

U A

U G

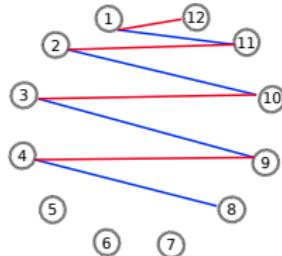
# Assignment of unconstraint paths

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

(( (( . . . ) ) ) )

NNNNNNNNNNNN

1 . . . 6 . . . 12



- Possible assignments for a base pair:

$$\sum_{i,j} BPT[i, j] = 6$$

- Possible assignments for a path of length two:

$$BPT = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \end{bmatrix}$$

$$\begin{pmatrix} & & \\ & & \\ \cdot & & \cdot \\ A & U & \end{pmatrix}$$

- Possible assignment for a path of length  $L$ :

$$C \quad G$$

$$\begin{matrix} G & C \\ G & U \end{matrix}$$

$$\begin{matrix} U & A \\ U & G \end{matrix}$$

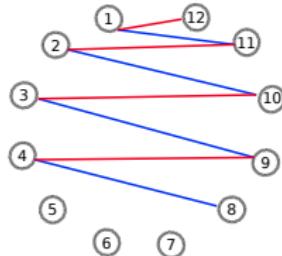
# Assignment of unconstraint paths

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

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

NNNNNNNNNNNN

1 . . . . 6 . . . . 12



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$$BPT = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \end{bmatrix}$$

(      )      .  
·      (      )  
A      U      A  
A      U      G  
C      G      C  
C      G      U  
G      C      G  
G      U      A  
G      U      G  
U      A      U  
U      G      U  
U      G      C

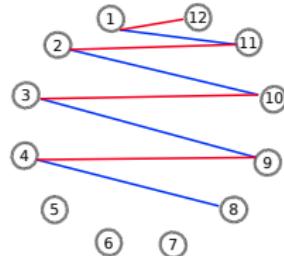
# Assignment of unconstraint paths

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

(( (( . . . ) ) ) )

NNNNNNNNNNNN

1 . . . 6 . . . 12



- Possible assignments for a base pair:

$$\sum_{i,j} BPT[i,j] = 6$$

- Possible assignments for a path of length two:

10

- Possible assignment for a path of length  $L$ :

$$BPT = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \end{bmatrix}$$

( ) .  
· ( )  
A U A  
A U G  
C G C  
C G U  
G C G  
G U A  
G U G  
U A U  
U G U  
U G C

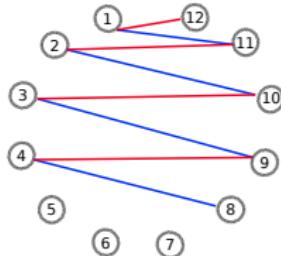
# Assignment of unconstraint paths

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

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

NNNNNNNNNNNN

1 . . . . 6 . . . . 12



- Possible assignments for a base pair:

$$\sum_{i,j} BPT[i, j] = 6$$

- Possible assignments for a path of length two:

10

- Possible assignment for a path of length  $L$ :

$$BPT = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \end{bmatrix}$$

$$BPT^2 = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 2 & 0 \\ 0 & 1 & 0 & 2 \end{bmatrix}$$

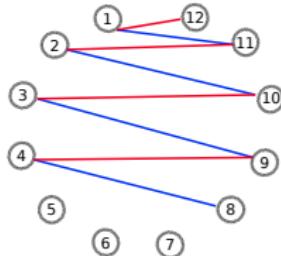
# Assignment of unconstraint paths

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

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

NNNNNNNNNNNN

1 . . . . 6 . . . . 12



- Possible assignments for a base pair:

$$\sum_{i,j} BPT[i,j] = 6$$

- Possible assignments for a path of length two:

$$\sum_{i,j} BPT^2[i,j] = 10$$

- Possible assignment for a path of length  $L$ :

$$BPT = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \end{bmatrix}$$

$$BPT^2 = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 2 & 0 \\ 0 & 1 & 0 & 2 \end{bmatrix}$$

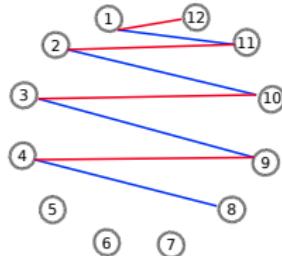
# Assignment of unconstraint paths

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$$BPT = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \end{bmatrix}$$

$$BPT^2 = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 2 & 0 \\ 0 & 1 & 0 & 2 \end{bmatrix}$$

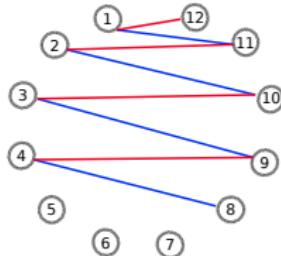
# Assignment of unconstraint paths

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NNNNNNNNNNNN

1 . . . . 6 . . . . 12



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$$\sum_{i,j} BPT[i,j] = 6$$

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$$\sum_{i,j} BPT^2[i,j] = 10$$

- Possible assignment for a path of length  $L$ :

$$\sum_{i,j} BPT^L[i,j]$$

$$BPT = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \end{bmatrix}$$

$$BPT^2 = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 2 & 0 \\ 0 & 1 & 0 & 2 \end{bmatrix}$$

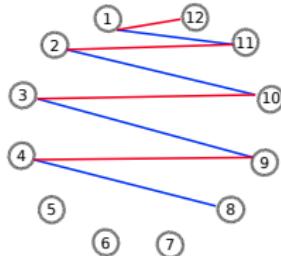
# Assignment of unconstraint paths

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

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

NNNNNNNNNNNN

1 . . . . 6 . . . . 12



$$4^3 \times \sum_{i,j} BPT^8[i,j]$$

- Possible assignments for a base pair:

$$\sum_{i,j} BPT[i,j] = 6$$

$$BPT = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \end{bmatrix}$$

- Possible assignments for a path of length two:

$$\sum_{i,j} BPT^2[i,j] = 10$$

$$BPT^2 = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 2 & 0 \\ 0 & 1 & 0 & 2 \end{bmatrix}$$

- Possible assignment for a path of length  $L$ :

$$\sum_{i,j} BPT^L[i,j]$$

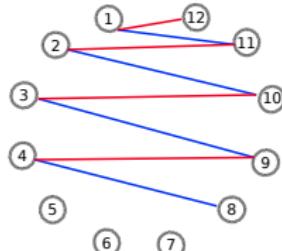
# Assignment of unconstraint paths

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

(( (( . . . ) ) ) )

NNNNNNNNNNNN

1 . . . 6 . . . 12



$$4^3 \times \sum_{i,j} BPT^8[i,j]$$

$$= 4^3 \times 178$$

$$= 11,392$$

- Possible assignments for a base pair:

$$\sum_{i,j} BPT[i,j] = 6$$

- Possible assignments for a path of length two:

$$\sum_{i,j} BPT^2[i,j] = 10$$

- Possible assignment for a path of length  $L$ :

$$\sum_{i,j} BPT^L[i,j]$$

$$BPT = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \end{bmatrix}$$

$$BPT^2 = \begin{bmatrix} 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 2 & 0 \\ 0 & 1 & 0 & 2 \end{bmatrix}$$

# Counting the correct way

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

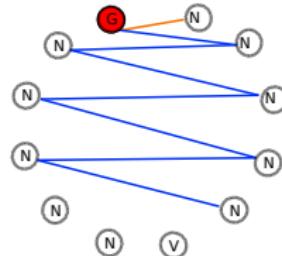
( ( ( ( . . . ) ) ) )

GNNNNNVNNNNN

Flamm, C., et al., 2001, Design of multistable RNA molecules, **RNA**, 7(2), 254-256

# Counting the correct way

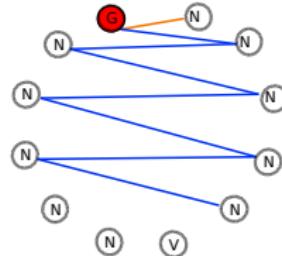
( (( ( ( . . . ) ) ) ) .  
( (( ( ( . . . ) ) ) )  
GNNNNNVNNNNN



Flamm, C., et al., 2001, Design of multistable RNA molecules, **RNA**, 7(2), 254-256

# Counting the correct way

( ( ( ( . . . ) ) ) ) .  
( ( ( ( . . . ) ) ) )  
GNNNNNVNNNNN

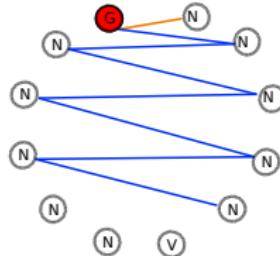


$$BPT^7 = \begin{bmatrix} 0 & 8 & 0 & 13 \\ 8 & 0 & 13 & 0 \\ 0 & 13 & 0 & 21 \\ 13 & 0 & 21 & 0 \end{bmatrix}$$

Flamm, C., et al., 2001, Design of multistable RNA molecules, **RNA**, 7(2), 254-256

# Counting the correct way

( ( ( ( . . . ) ) ) ) .  
( ( ( ( . . . . ) ) ) )  
GNNNNNVNNNNN

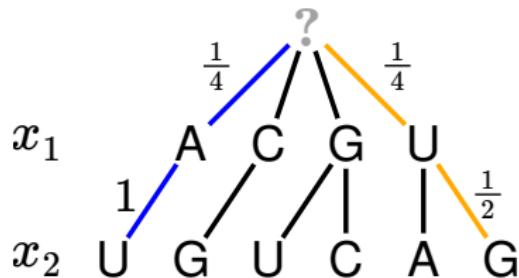


$$BPT^7 = \begin{bmatrix} 0 & 8 & 0 & 13 \\ 8 & 0 & 13 & 0 \\ 0 & 13 & 0 & 21 \\ 13 & 0 & 21 & 0 \end{bmatrix}$$

$$2 \times (13 + 21) \times 4^2 \times 3 = 3,264$$

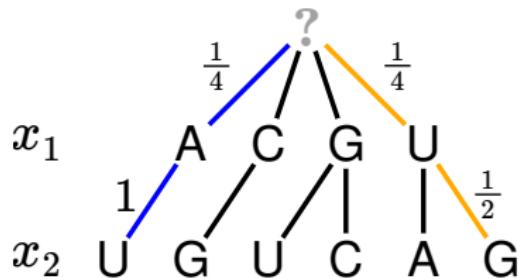
Flamm, C., et al., 2001, Design of multistable RNA molecules, **RNA**, 7(2), 254-256

# Uniform Sampling of a Base Pair



Flamm, C., et al., 2001, Design of multistable RNA molecules, **RNA**, 7(2), 254-256

# Uniform Sampling of a Base Pair



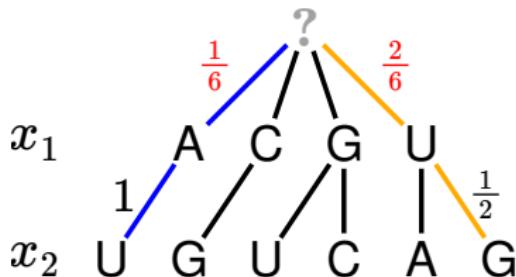
uniform decision  $\neq$  uniform sampling

$$P(AU) = \frac{1}{4} \times 1 = \frac{1}{4}$$

$$P(UG) = \frac{1}{4} \times \frac{1}{2} = \frac{1}{8}$$

Flamm, C., et al., 2001, Design of multistable RNA molecules, **RNA**, 7(2), 254-256

# Uniform Sampling of a Base Pair



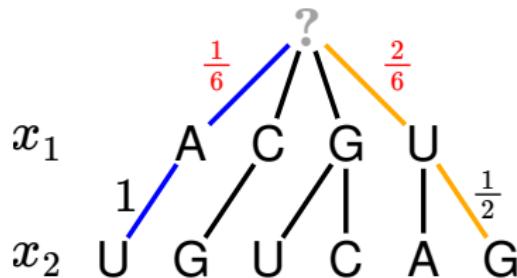
uniform decision  $\neq$  uniform sampling

$$P(AU) = \frac{1}{4} \times 1 = \frac{1}{4}$$

$$P(UG) = \frac{1}{4} \times \frac{1}{2} = \frac{1}{8}$$

Flamm, C., et al., 2001, Design of multistable RNA molecules, **RNA**, 7(2), 254-256

# Uniform Sampling of a Base Pair



**weighted decision = uniform sampling**

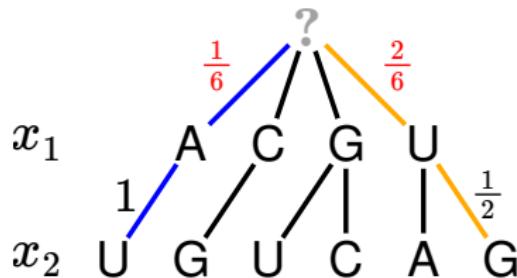
$$P(x_1 = U) = \frac{N(\{x_1, x_2\} | x_1 = U)}{N(x_1, x_2)} = \frac{2}{6}$$

$$P(AU) = \frac{1}{6} \times 1 = \frac{1}{6}$$

$$P(UG) = \frac{2}{6} \times \frac{1}{2} = \frac{1}{6}$$

Flamm, C., et al., 2001, Design of multistable RNA molecules, **RNA**, 7(2), 254-256

# Uniform Sampling of a Base Pair



**weighted decision = uniform sampling**

$$P(x_1 = \textcolor{orange}{U}) = \frac{N(\{x_1, x_2\} | x_1 = \textcolor{orange}{U})}{N(x_1, x_2)} = \frac{2}{6}$$

$$P(\textcolor{blue}{AU}) = \frac{1}{6} \times 1 = \frac{1}{6}$$

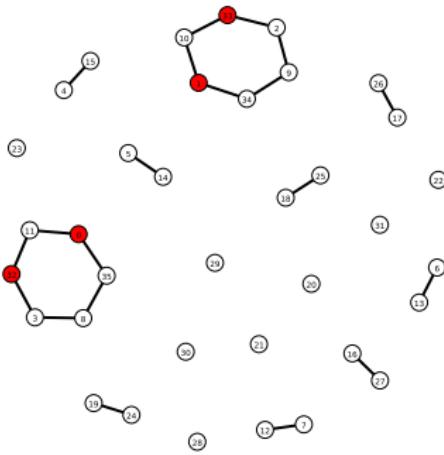
$$P(\textcolor{orange}{UG}) = \frac{2}{6} \times \frac{1}{2} = \frac{1}{6}$$

Flamm, C., et al., 2001, Design of multistable RNA molecules, **RNA**, 7(2), 254-256

# Example: 35nt long and three structural states

```
(( (( (....) ) ) ) .... (( (( (....) ) ) ) ....  
..... (( (( (.... (( (( (....) ) ) ) .... ) ) )  
(( ((((( (....) ) ) ) (( (( (....) ) ) ) .... ) ) )
```

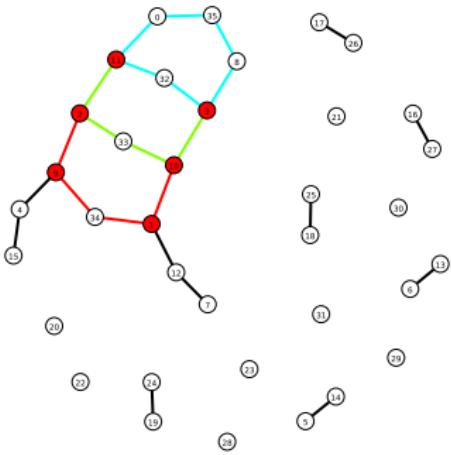
Size of solution space: 1.42658e+14



# Example: 35nt long and four structural states

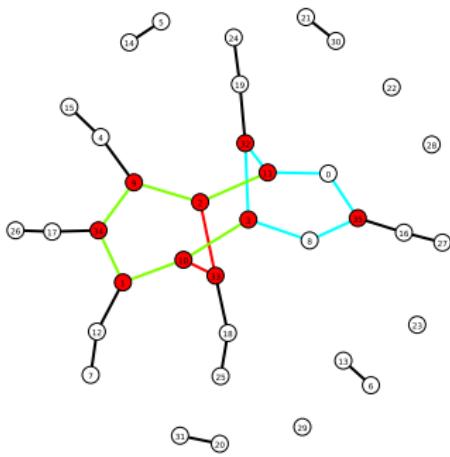
```
(( (( (....) ) ) ) .... (( (( (....) ) ) ) ....  
..... (( (( (.... (( (( (....) ) ) ) .... ) ) )  
(( ((((( (....) ) ) ) (( (( (....) ) ) ) .... ) ) )  
. (( (( (....) ) ) ) .... (( (( (....) ) ) ) ....
```

Size of solution space: 1.24018e+13



# Example: 35nt long and five structural states

```
(((((.....)))).....((((.....)))).....  
.....(((((.....(((((.....)))).....))))  
(((((((((.....))))(((((.....)))).....))))  
.((((.....))))....((((.....)))).....  
.((((.....))))....((((((.....)))))))  
Size of solution space: 7.08853e+10
```



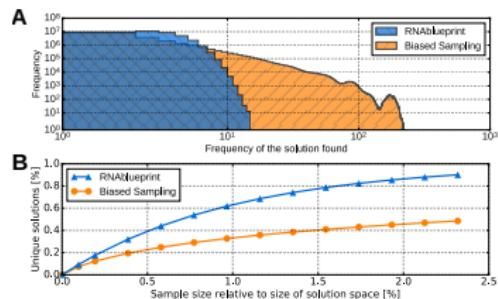
# Integrated Design Tool



## RNAblueprint

A tool that implements a graph coloring approach to **sample nucleic acid sequences** ...

- compatible to **multiple structural and sequence constraints**
- **fairly drawn** from the whole solution space



<https://github.com/ViennaRNA/RNAblueprint>

Hammer, S., et al., 2017, RNAblueprint: flexible multiple target nucleic acid sequence design. *Bioinformatics* 33(18): 2850-58.

# Objective Function

- Not all compatible sequences are equally good when evaluated under certain assumptions.
- Depending on the application the objective function needs to be developed.

# Objective Function

- Not all compatible sequences are equally good when evaluated under certain assumptions.

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

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

GNNNNNVNNNNN

$$f(x) = \underbrace{((EOS_1 + EOS_2) - 2EFE)}_{\text{dominate ensemble of structures}} + \underbrace{\xi |EOS_1 - EOS_2|}_{\text{equally stable}}$$

$$f(\text{GGGAGCCCC}) = 3.82 \quad f(\text{GAAUCAUUU}) = 8.42$$

$$f(\text{GGGUCCCCU}) = 4.71 \quad f(\text{GGGGAUUCU}) = 5.71$$

- Depending on the application the objective function needs to be developed.

# Objective Function

- Not all compatible sequences are equally good when evaluated under certain assumptions.

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

(( (( . . . ) ) ) )

GNNNNNVNNNNN

$$f(x) = \underbrace{((EOS_1 + EOS_2) - 2EFE)}_{\text{dominate ensemble of structures}} + \underbrace{\xi |EOS_1 - EOS_2|}_{\text{equally stable}}$$

$$f(\text{GGGAGCCCC}) = 3.82 \quad f(\text{GAAUCAUUU}) = 8.42$$

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- Depending on the application the objective function needs to be developed.

# RNA design aims on the reverse

Sequence  $\Leftarrow$  Structure  $\Leftarrow$  Function

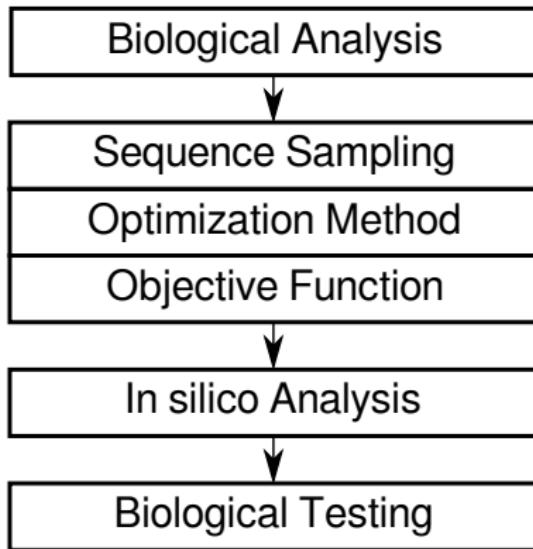
- ① Which sequence optimally folds into given target structure(s)?
- ② How to implement novel functions?

# RNA design aims on the reverse

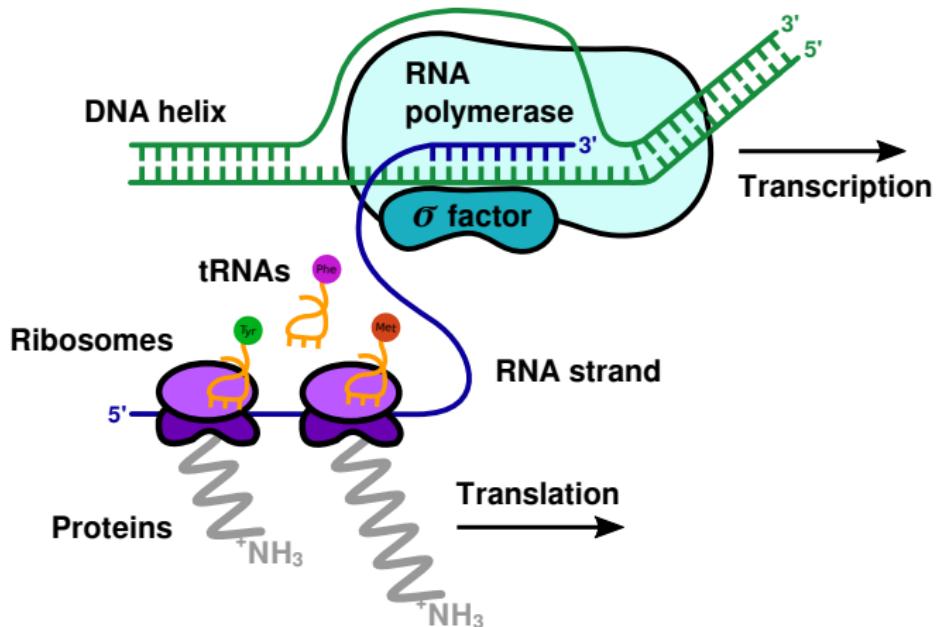
Sequence  $\Leftarrow$  Structure  $\Leftarrow$  Function

- ① Which sequence optimally folds into given target structure(s)?
- ② How to implement novel functions?

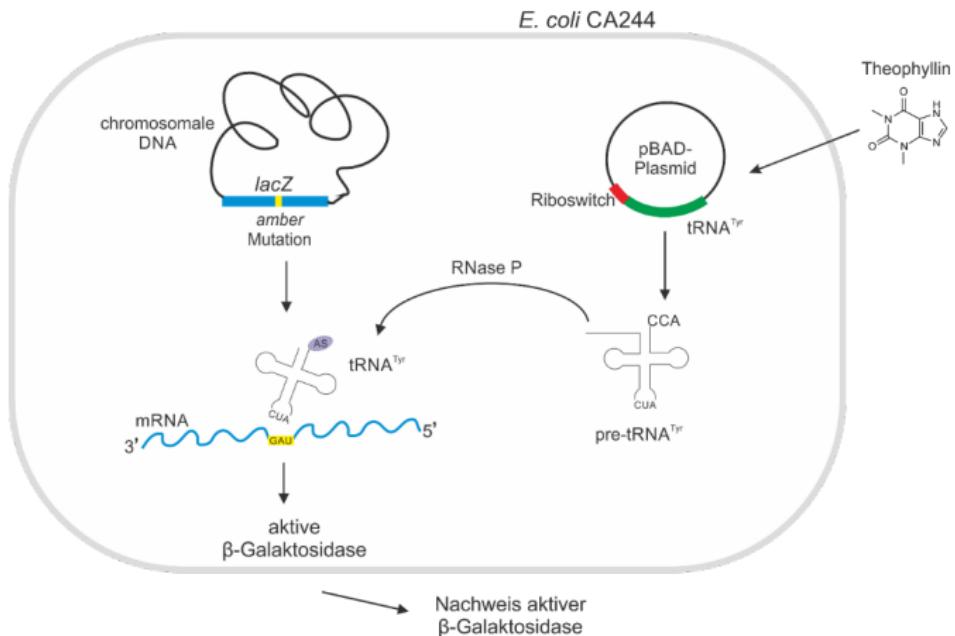
# Design Pipeline



# Transcription/Translation

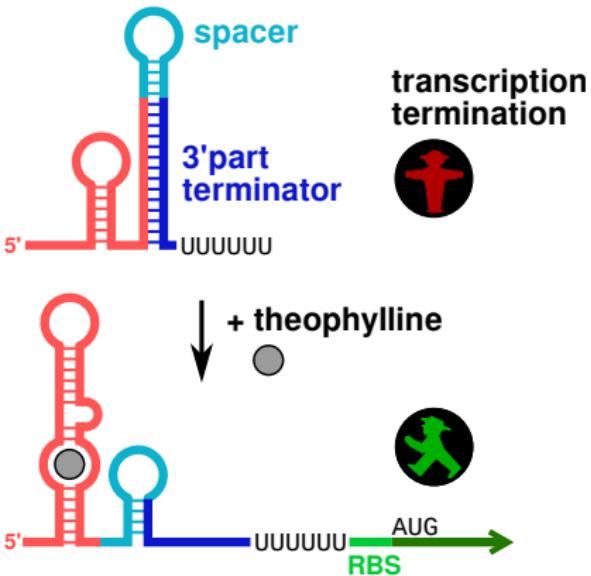
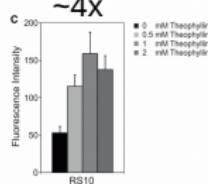
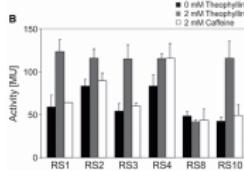
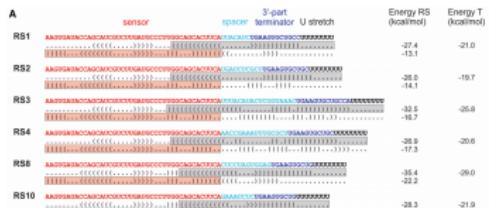


# tRNA processing mechanism

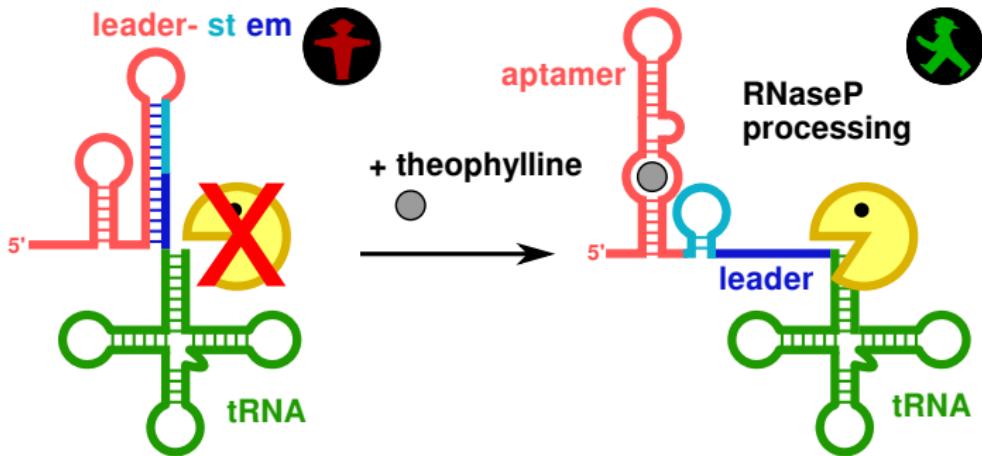


# Artificial Riboswitch

- Theophylline triggered
  - Transcriptional
  - Reporter Gene



# RNaseP riboswitch idea



# Copy&Paste Design

## construct I (RNaseP RS, R273)

> Theophylline RS w/o poly U fused to su3+

AAGUGAUACCAGCAUCGCUUGAUGCCCUUGGCAGCACUUCA GAAAUUCUCAAGUGCUGCCAAGGUGGGGUUCC  
CGAGCGGCCAAAGGGAGCAGACUCUAAAUCUGCCGUCAUCGACUUCGAAGGUUCGAUCCUCCCCCACCACCA

cleavage  
↓

tRNA spacer aptamer

- This construct is without the original 5'-leader sequence of the tRNA

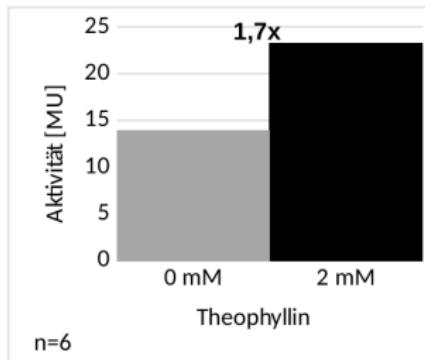
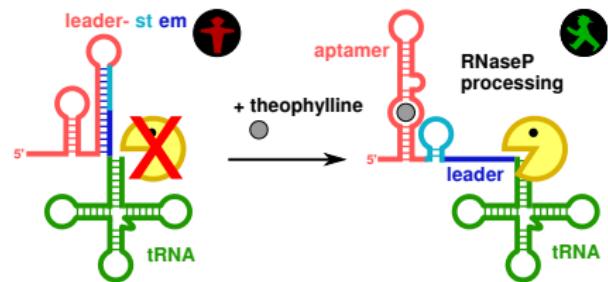
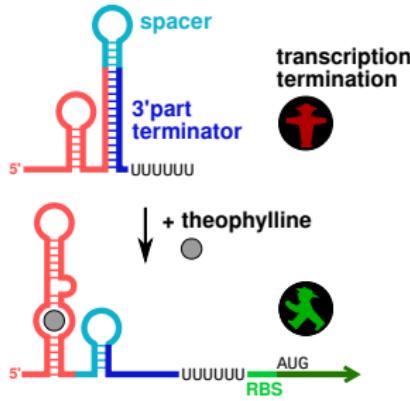
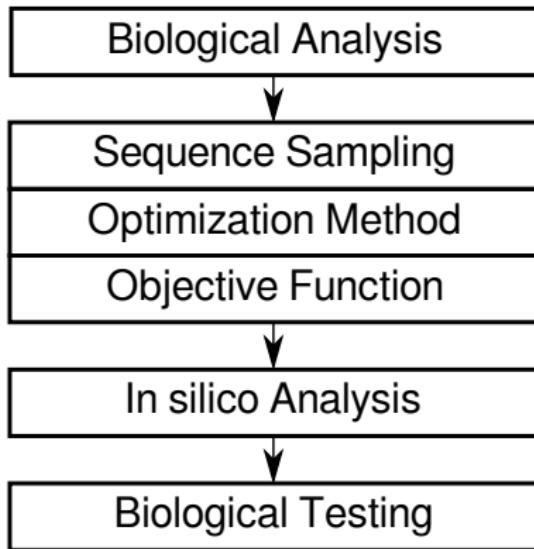


Figure 1: ONPG- assay with construcht I (RNase R RS).

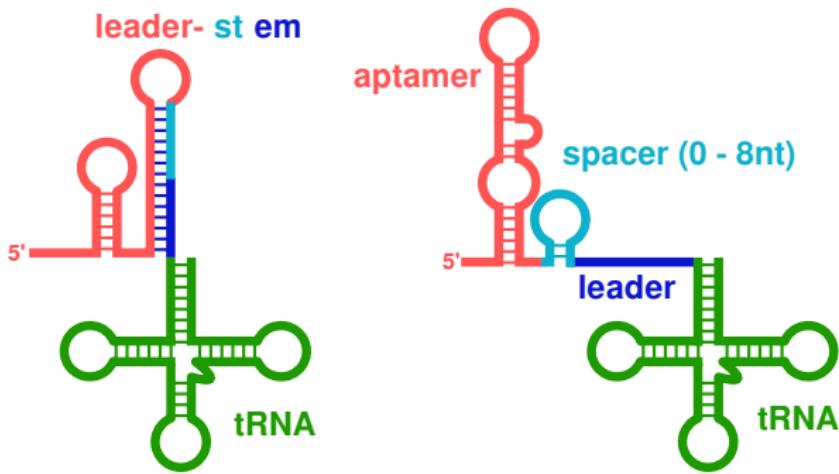
# Timing is important!



# Design Pipeline



# Constrained Sampling



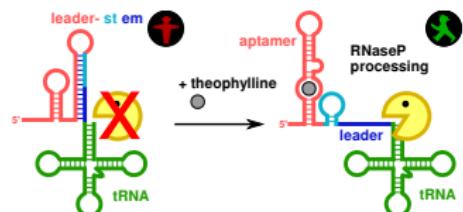
Sequence + Structure constraints →



# Objective Functions

- Accessibility score [0,1]:

$$f(x) = P_{x,C_{theo}}(\Theta_{leader}) \cdot (1 - P_x(\Theta_{leader}))$$



- State score [0,1]:

$$f(x) = \underbrace{P_x(\Theta_{leader-stem})}_{\text{leader-stem} \rightarrow 1} \cdot \underbrace{\left( \frac{1 - P_x(\Theta_{aptamer}) + P_x(\Theta_{leader-stem})}{2} \right)^2}_{\text{difference to aptamer} \rightarrow 1}$$
$$\cdot \underbrace{P_{x,C_{theo}}(\Theta_{aptamer})}_{\text{aptamer} \rightarrow 1} \cdot \underbrace{\left( \frac{1 - P_{x,C_{theo}}(\Theta_{leader-stem}) + P_{x,C_{theo}}(\Theta_{aptamer})}{2} \right)^2}_{\text{difference to leader-stem} \rightarrow 1}$$

$x \dots$  Sequence

$\Theta \dots$  Structure

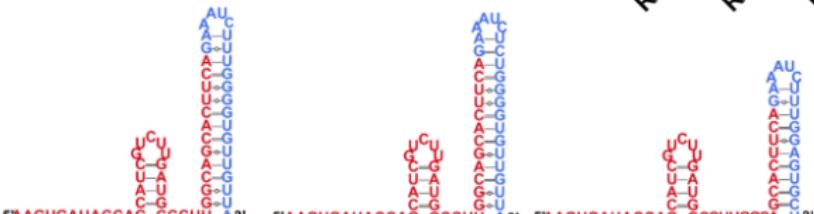
$C_{theo} \dots$  Soft-Constraint for Theophylline

# Laboratory Analysis



Theo-RP-RS1

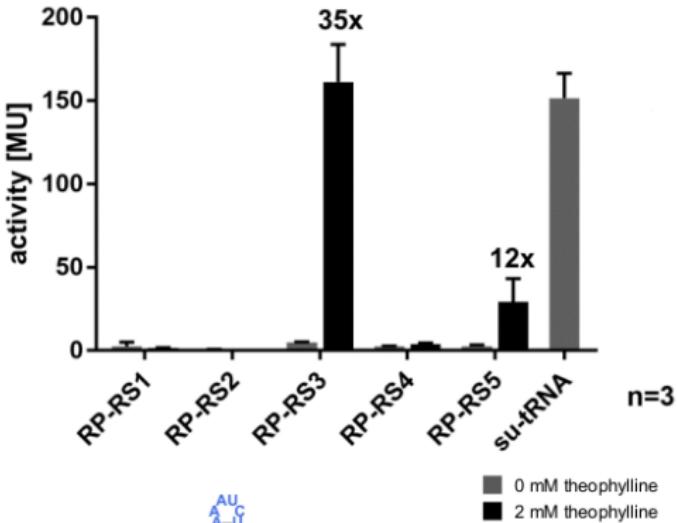
Theo-RP-RS2



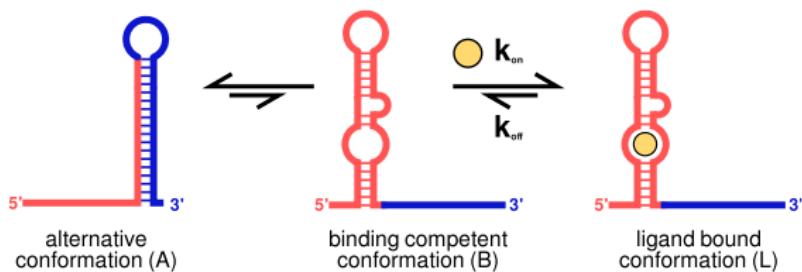
Theo-RP-RS3

Theo-RP-RS4

Theo-RP-RS5



## LIGAND DEPENDENT RNA SWITCH



Findei $\beta$ , S., Hammer, S., Wolfinger, M. T., Kühnl, F., Flamm, C., & Hofacker, I. L. (2018). In silico design of ligand triggered RNA switches. Methods.

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