# Modeling RNA-RNA interaction formation on direct paths

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kinetic models

- explore full structure space
- coarse graining
- computationaly costly

thermodynamic models

- established tools
- efficient
- no kinetics

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 $\Rightarrow$  thermodynamic structure candidate

Kinetically favorable path to forming this structure?

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biological relevant vs. non-functional interactions

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biological relevant vs. non-functional interactions kinetc features for improve interaction predictions















 $\Rightarrow$  structures on direct paths



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- fixed set of base pairs
- only consecutive substructures
- move set: base pair opening and closing

# Free energy of interaction structures



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#### intramolecular folding

fast (instant) unpaired probability of interaction sites

slow (minimal) removal of conflict base pairs









3' sRNA

5' mRNA



- seed stability
- seed accessibility
- Iocal minima

. . .

barriers on paths

## DP algorithm for minimal folding barriers

$$B_{s}(i,j) = \min \begin{cases} \max \begin{cases} E(i,j) & \text{if } i < s \\ B_{s}(i+1,j) & \\ \max \begin{cases} E(i,j) & \\ B_{s}(i,j-1) & \\ E(s,s) & \text{if } s = i = j \end{cases}$$

$$E(i,j) = E^{\text{hybrid}}(i,j) + E^{\text{unpaired}}(i,j)$$
$$E^{\text{hybrid}}(i,j) = \min \begin{cases} E^{\text{hybrid}}(i+1,j) + L(i,i+1) & \text{if } i < j \\ E_{\text{init}} & \text{if } i = j \end{cases}$$

$$\frac{P_i(t)}{dt} = \sum_{i \neq j} (P_j(t)k_{ji} - P_i(t)k_{ij})$$

- set of structures  $\Rightarrow$  states  $\Omega$
- move set M $\Rightarrow$  neigborhood relation
- energy function E
- free energy difference  $\Delta G^{\ddagger}$  $\Rightarrow$  folding rate  $k_{ij}$

$$k_{ij} = egin{cases} k_0 ext{ if } \Delta G^{\ddagger} \leq 0, \ k_0 e^{rac{-\Delta G^{\ddagger}}{RT}} ext{ otherwise} \end{cases}$$



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## Do kinetic feature provide new information?



(72 sRNA + mRNA pairs from E. coli )

Increased prediction performance with kinetics

E. coli data set:

- ▶ native sRNA + native mRNA (104 pairs)
- ▶ native sRNA + 4 × shuffled mRNA (420 pairs)
- $\Rightarrow$  thermodynamic interaction prediction
- $\Rightarrow$  compute kinetic features
- $\Rightarrow$  train ML classifier

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#### linear discriminator, 10 fold validation

features	MCC	recall	precision
$E_{ m therm}$	0.53	0.50	0.76
$E_{\rm therm}$ + seed	0.60	0.52	0.86
$E_{\rm therm}$ + detailed dynamics	0.61	0.58	0.81
$E_{\rm therm}$ + seed + detailed dynamics	0.65	0.65	0.84
$E_{\rm therm}$ + seed + detailed dynamics + barrier	0.71	0.67	0.87

no overlap in sRNAs between test and training set

#### Summary

- computational model for interaction formation
- efficient computation methods for features
- benchmark prediction capabilities of features
- ► test **mechanistic** hypothesises
- $\Rightarrow$  stable seed interaction
- $\Rightarrow$  fast formation of full interaction

Outlook

easy to use prediction tool

#### In collaboration with ...

#### Vienna

- ► Ivo L. Hofacker
- ► Irene K. Beckmann
- ▶ and the TBI Team

Paris

Sebastian Will

#### Freiburg

- Rolf Backofen
- Martin Raden

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## seed accessibility and folding barrier example



Cost for keeping the pairing positions of the interaction unpaired in the intramolecular structure

#### Dsra mechanism



#### Example energy landscape (E. coli: DsrA rpoS)

Free energy AG in kcal/mol for each interaction substructure from base pair i to base pair j

