

Tractable Kinetics of RNA–Ligand Interaction

On the Folding Kinetics of Riboswitches

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RNA structure, design and interaction with proteins
Mini-symposium at JOBIM 2020

July 1, 2020

Simulating RNA Folding

RNA sequence \rightarrow (secondary) structure(s) \rightarrow function

Types of folding simulations:

- thermodynamic:
 - ▶ structure probabilities
in equilibrium
 - ▶ fast
(use e. g. ViennaRNA)
- kinetic:
 - ▶ yields distribution of
structures over time
 - ▶ computationally harder,
size restrictions
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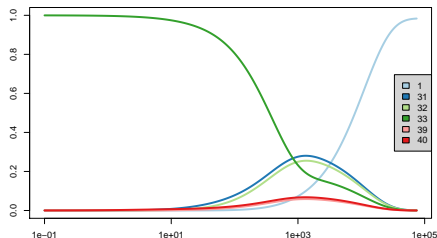
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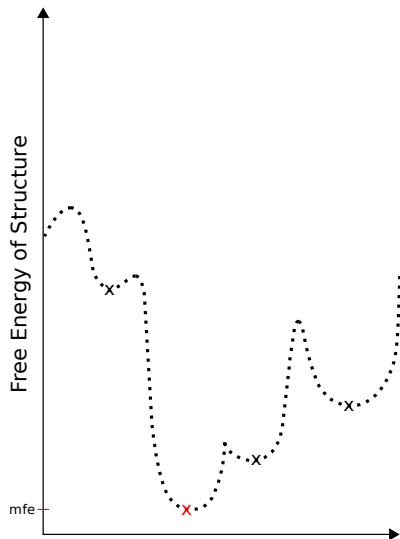
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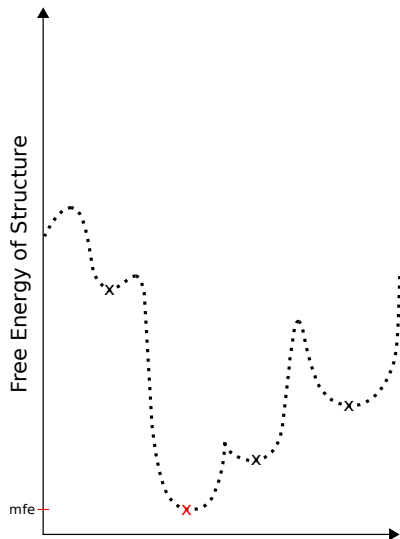
RNA Folding Kinetics: Energy Landscapes

- goal: model folding process
- structure $\varphi \mapsto$ energy value E_φ (TURNER energy model)
- *transitions* between structures
 - ▶ insertion, deletion, shift
 - ▶ only if distance is 1
 - ▶ rate constants $r_{\chi \leftarrow \varphi}(E_\varphi, E_\chi)$
- $\#\text{structures}(s) \in \mathcal{O}(2.62^{|s|})$
 - ▶ coarse graining necessary
 - ▶ *gradient basins* as macrostates
 - ▶ *macrorate* constants $r_{j \leftarrow i}$



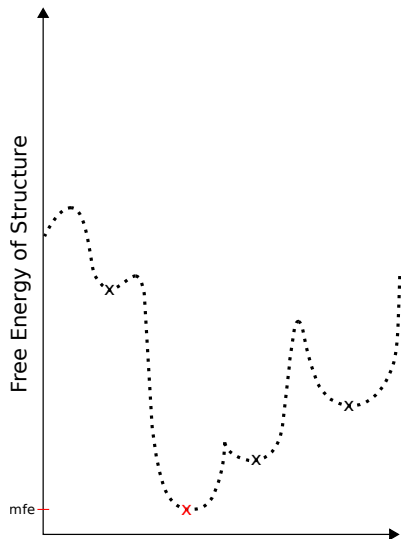
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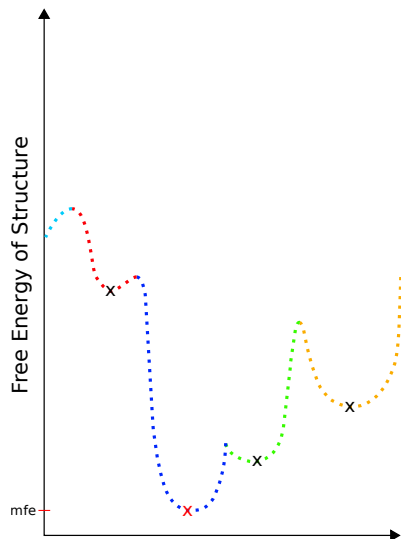
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Rate Coefficients

- general form (ARRHENIUS equation):

$$r_{\chi \leftarrow \varphi} = a_{\text{RNA}} \cdot \exp\left(-\frac{E_a(\chi \leftarrow \varphi)}{RT}\right)$$

for RNA structures φ and χ

(E_a – activation energy R – gas constant T – temperature)

- a_{RNA} – “collision rate” (from experiment)
- activation energy (METROPOLIS rule):

$$E_a(\chi \leftarrow \varphi) = \begin{cases} \max(E_\varphi, E_\chi) - E_\varphi & \text{if } \varphi \text{ and } \chi \text{ are neighbors} \\ \infty & \text{else} \end{cases}$$

- macrorates for basins i, j (WOLFINGER et al.):

$$r_{j \leftarrow i} = \sum_{\substack{\varphi \in i \\ \chi \in j}} \underbrace{\Pr[\varphi | i]}_{= \frac{z(\varphi)}{z(i)}} \cdot r_{\chi \leftarrow \varphi}$$

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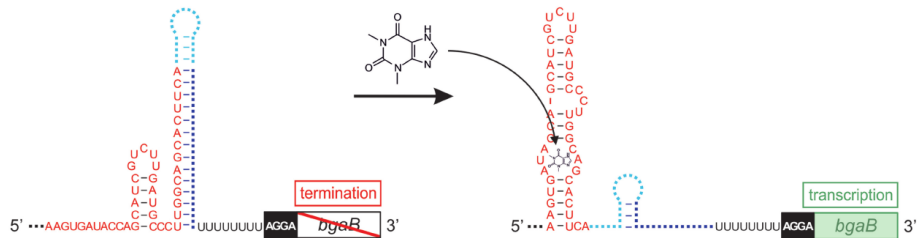
Riboswitches ...

- ... are regulatory RNA elements in 5'-UTR of genes
- regulate transcription or translation depending on a *ligand*

- consist of ...
 - ▶ an *aptamer*: contains *binding pocket* for ligand L
 - ▶ an *actuator*, e. g. a *terminator hairpin*
- binding of ligand yields *energy bonus* $\theta_L < 0$
 $\implies E_{L\varphi} = E_\varphi + \theta_L$
- time-critical interaction with RNA polymerase / ribosome

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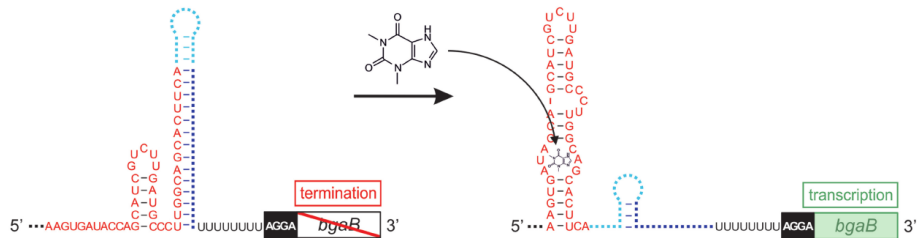
transcriptional, theophylline-controlled ON switch

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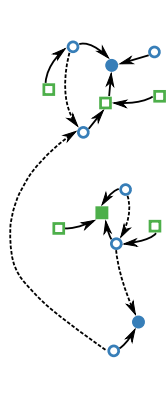


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Riboswitch Energy Landscapes



monomers

dimers



(R – RNA

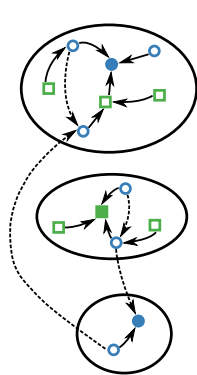


L – ligand



LR – dimer)

Riboswitch Energy Landscapes



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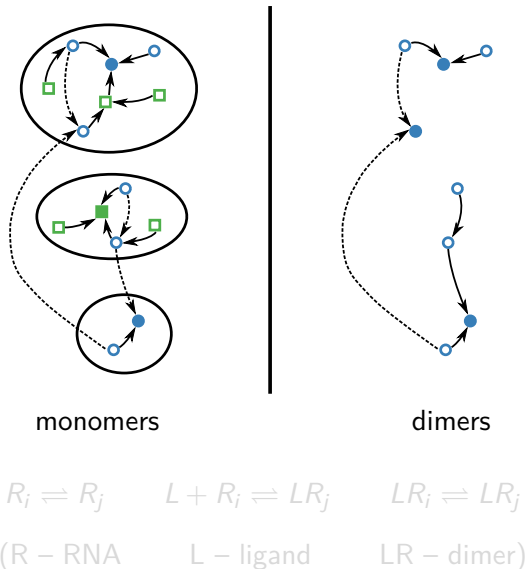


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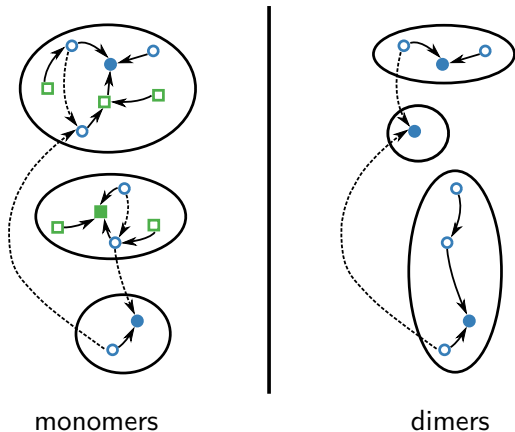
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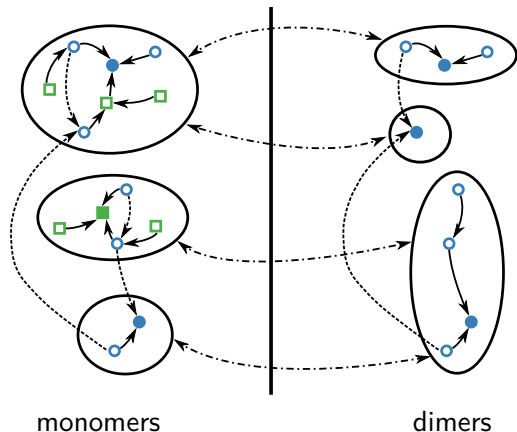


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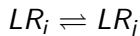
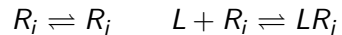
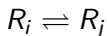
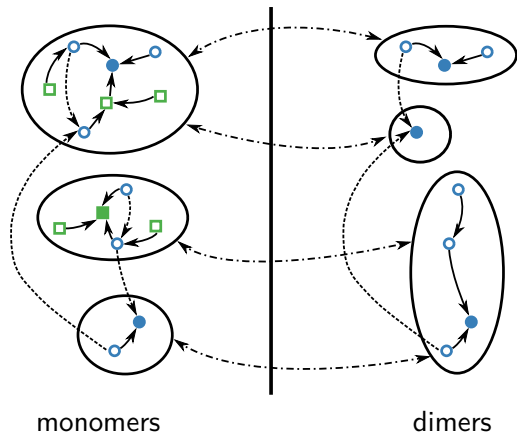


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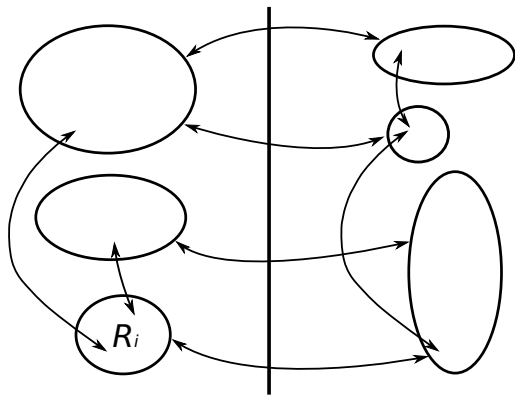


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L – ligand

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Modelling the System



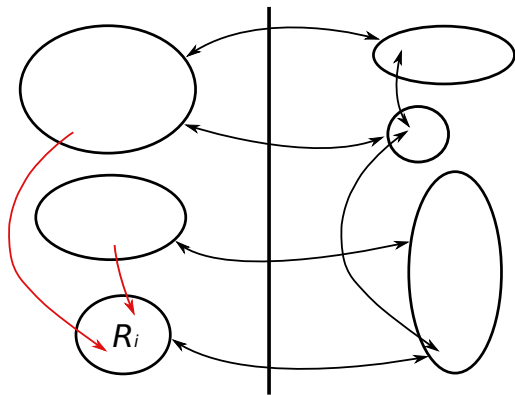
By first-order rate laws:

(assumption: $[L] = l_0 = \text{const}$)

$$[\dot{R}_i] = + \sum_{\substack{1 \leq k \leq n \\ k \neq i}} r_{i \leftarrow k} [R_k] + \sum_{1 \leq k \leq m} r_{i \leftarrow Lk} [LR_k] - \sum_{\substack{1 \leq k \leq n \\ k \neq i}} r_{k \leftarrow i} [R_i] - \sum_{1 \leq k \leq m} r_{Lk \leftarrow i} \cdot l_0 \cdot [R_i]$$

for n monomer and m dimer basins.

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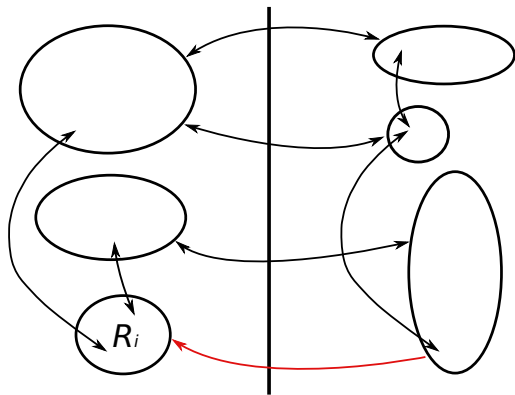
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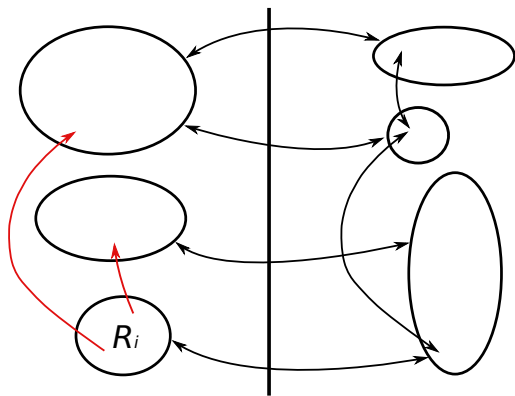
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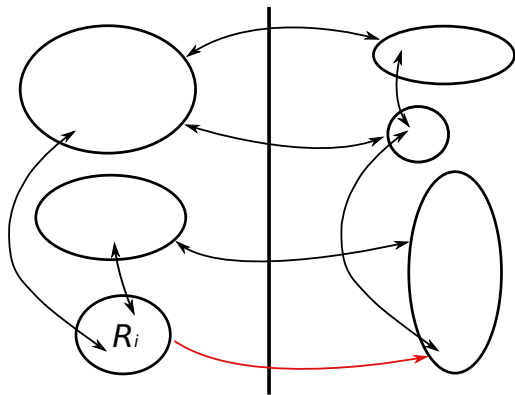
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Modelling the System ctd.

By first-order rate laws:

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$$(i = 1, \dots, n) \quad [\dot{R}_i] = + \sum_{\substack{1 \leq k \leq n \\ k \neq i}} r_{i \leftarrow k} [R_k] \quad + \sum_{1 \leq k \leq m} r_{i \leftarrow Lk} [LR_k] \\ - \sum_{\substack{1 \leq k \leq n \\ k \neq i}} r_{k \leftarrow i} [R_i] \quad - \sum_{1 \leq k \leq m} r_{Lk \leftarrow i} \cdot l_0 \cdot [R_i],$$

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for n monomer and m dimer basins.

Simple, isn't it?

After some clever rewriting ...

$$\dot{x} = R(l_0) \cdot x$$

with ...

- $x = ([R_1], \dots, [R_n], [LR_1], \dots, [LR_m])^T$ – vector of concentrations
- $R(l_0)$ – (constant) rate coefficient matrix, more detailed:

$$R(l_0) = \begin{pmatrix} A & C \\ l_0 \cdot D & B \end{pmatrix}$$

where

- ▶ A – monomer rate coefficients, $(a_{ji}) = r_{j \leftarrow i}$
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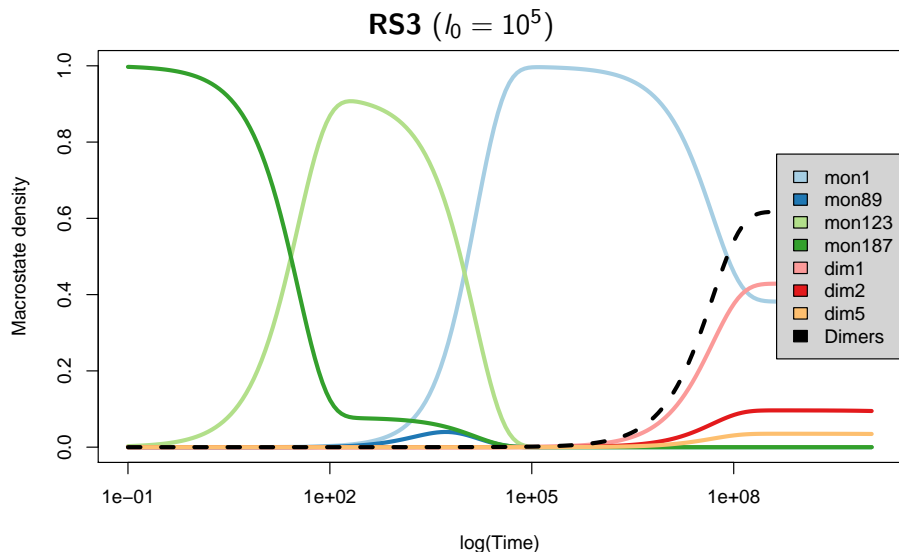
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Kinetics of Theophylline Riboswitch RS3

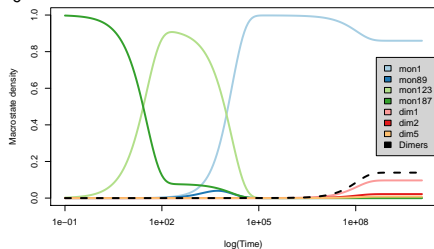
RNASubopt \rightarrow barriers + “monomer \leftrightarrow dimer” rates \rightarrow treekin



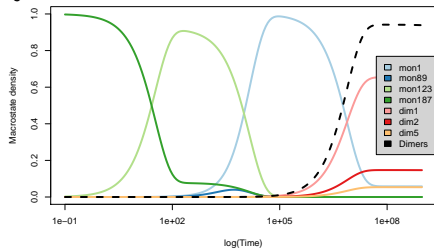
Concentration-dependent Behavior

RS3

$I_0 = 10^4$



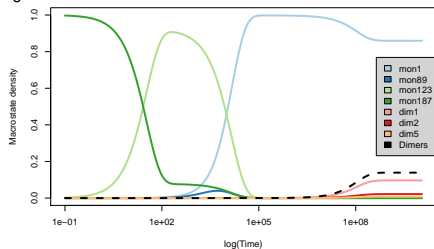
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Concentration-dependent Behavior

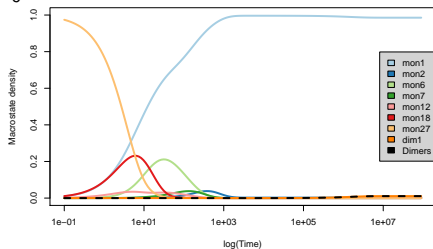
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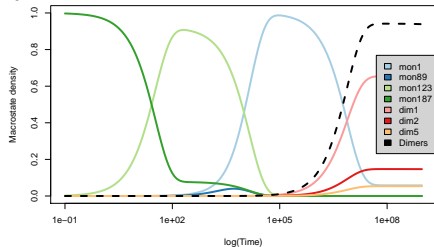


RS3 shortened

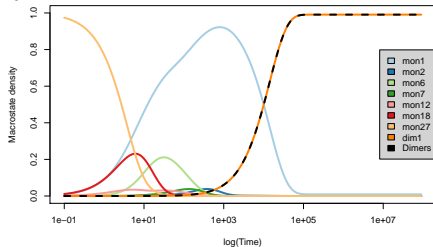
$l_0 = 10^{-8}$



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Cotranscriptional Folding

BarMap constructs mappings for a sequence of coarse-grained energy landscapes and performs a kinetic folding simulation on them. [Hofacker et al. 2010]
Applications: cotranscriptional folding, RNA thermometers, ...

Extension of barriers and BarMap to include the presented interaction model are available. [Wolfinger et al. 2018]

BarMap-QA: ready-to-use pipeline utilizing BarMap for co-transcriptional folding (quality control statistics, result visualizations, Docker container)

[Kühnl et al. 2020], caveat: no ligand support yet

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- **fast** re-computation of **results** for arbitrary ...
 - ▶ ... initial RNA distributions
 - ▶ ... ligand concentrations
 - ▶ ... pre-exponential factors
- enables *in silico* **testing of designed riboswitches**
- can be **refined** as more precise measurements of **collision and folding rates** become available

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Additional Material

Rate Coefficients in Riboswitch Landscapes

Lemma

- ① coefficient for $LR_i \rightarrow LR_j$ (dimer-dimer rate):

$$r_{L\chi \leftarrow L\varphi} = r_{\chi \leftarrow \varphi} \quad \Longrightarrow \quad r_{Lj \leftarrow Li} = r_{j \leftarrow i}$$

- ② coefficient for $L + R_i \rightarrow LR_j$ (dimerization rate):

$$r_{L\varphi \leftarrow \varphi} = a_{dim} \quad \Longrightarrow \quad r_{Lj \leftarrow i} = a_{dim} \cdot \frac{Z[i \cap j]}{Z[i]}$$

- ③ coefficients for $LR_i \rightarrow L + R_j$ (dissociation rate):

$$r_{\varphi \leftarrow L\varphi} = a_{dim} \cdot \exp\left(\frac{\theta_L}{RT}\right) \quad \Longrightarrow \quad r_{j \leftarrow Li} = a_{dim} \cdot \frac{Z[i \cap j]}{Z[i]} \cdot \exp\left(\frac{\theta_L}{RT}\right)$$

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