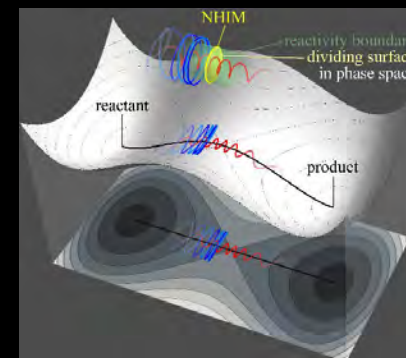


A Structure of Kinetic Hierarchy in Reaction Network



Tamiki Komatsuzaki

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Adjunct, The Institute of Scientific and Industrial Research, Osaka University

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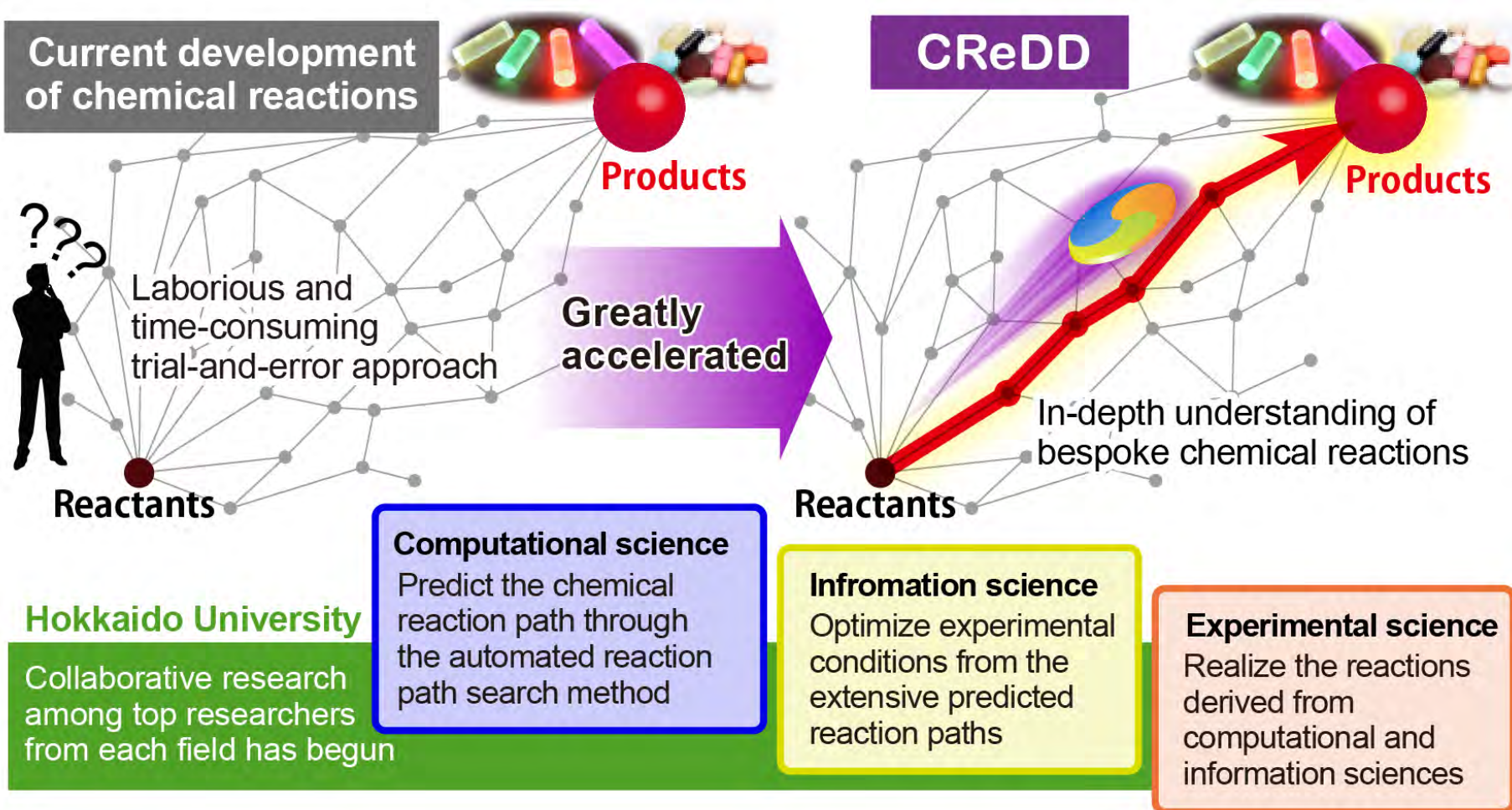
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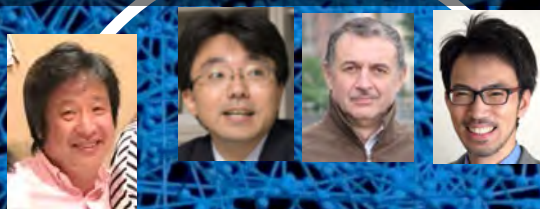


Institute for Chemical Reaction Design and Discovery

<https://www.icredd.hokudai.ac.jp/>

Information Science

AI, Machine Learning, Network theory,
Information Theory, Applied Mathematics



Computational Science

Quantum Chemistry
MD simulation

Experimental Science

Synthetic Chemistry,
Biology, Medicine









ICReDD Organization of 2022




























































- Calculation
- Information
- Experiment
- Administration

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1. Background and Motivation
2. Transition Disconnectivity Graph
3. Network Transition States
4. Lumping Theory to Preserve Hierarchical Timescales
5. Outlooks

Lagrangian Coherent Structure

H. Teramoto et al., *Chaos* **23**, 043107 (2013)
P. Nag et al. *J. Chem. Phys.* **141**, 104907(2014)

Higher rank saddles

Y. Nagahata et al. *Phys. Rev. E*
87, 62817 (2013); *ibid.* **88**, 42923 (2013)

Bifurcation of NHIM

C.-B. Li, T. Komatsuzaki, *J. Chem. Phys.* **130**, 124116, (2009)

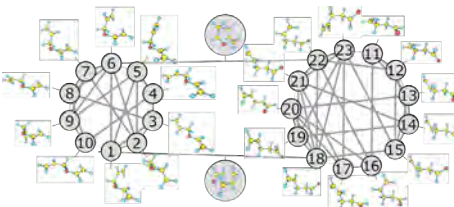


Origin of Stochasticity in Transitions

Komatsuzaki et al. *PNAS* **98** 7666 (2002), C.-B. Li et al., *PRL* **97**, 028302 (2006); H. Teramoto et al., *PRL* **106**, 054101 (2011); **115**, 093003 (2015); *Nonlinearity* **28**, 2677 (2015); *PRL*, 054101 (2015)

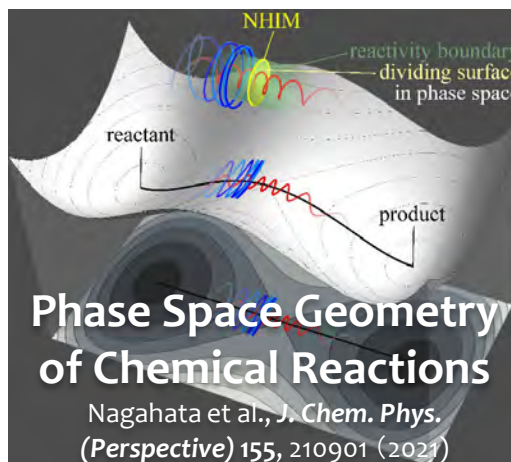
TS in Complex Network

Y. Nagahata, S. Maeda, H. Teramoto, T. Horiyama, T. Taketsugu, T. Komatsuzaki et al. *JPCB* **120**(8), 1961-1971 (2015).



multidimensional Energy Landscape Based on single molecule time series

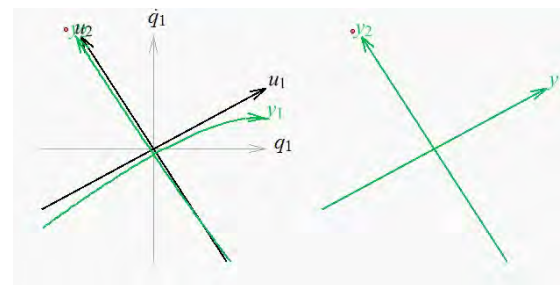
A. Baba et al., *PNAS* **104**, 19297 (2007); *PCCP*, **13**, 1395 (2011); K. Kamagata et al., *JACS* **134**, 11525 (2012); J.N. Taylor et al., *Sci. Rep.* **5**, 9174 (2015)
J. Alfermann et al., *Nature Chem. Bio.* **13**, 1009 (2017)



Phase Space in Wells

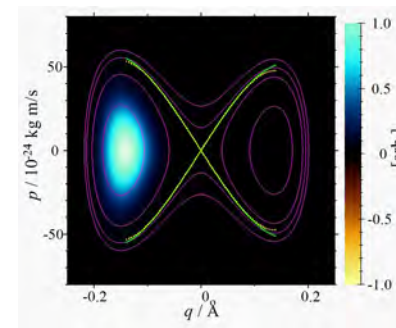
A. Shojiguchi et al. *Phys. Rev. E* **76**, 056205 (2007); *ibid.* **75**, 035204(R) (2007); H. Teramoto et al. *J. Chem. Phys.* **129**, 094302 (2008); *Phys. Rev. E* **78**, 017202 (2008); *Theor. Chem. Acc.* **133**, 1571, (2014)(invited); Y. Mizuno et al., *Physica D* **428**, (2021) (invited)

No-return TS in Dissipative systems



S. Kawai, T. Komatsuzaki, *Phys. Chem. Chem. Phys. (Perspective)* **13**, 21217 (2011); *Phys. Chem. Chem. Phys.* **12**, 15382 (2010); *ibid.* **12**, 7626 (2010); *ibid.* **12**, 7636 (2010); *J. Chem. Phys.* **134**, 114523, (2011); *ibid.* **131**, 224505 (2009) *ibid.* **131**, 224506 (2009); Nagahata et al., *J. Chem. Phys. (Perspective)* **155**, 210901 (2021) (invited) (selected as cover art)

Optimal control theory in chemical reactions

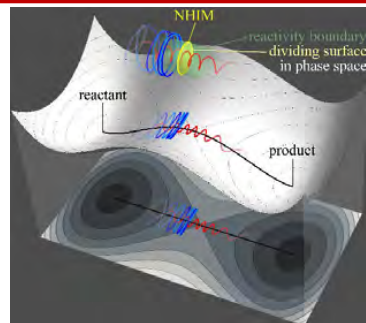


S. Kawai, T. Komatsuzaki, *Bull. Chem. Soc. Japan* **85**, 854 (2012); *J. Chem. Phys.* **134**, 024317 (2011)

To understand chemical reactions including structural transitions

1 : Describe the “space/scaffold” where chemical reactions take place (static)

- ✓ Electronic structure theories (e.g., HOMO-LUMO theory, QM/MM)
- ✓ Energy landscape or reaction (conformation) network representation



2 : Understand why and how fast reaction takes place (kinetic, dynamic)

- ✓ Transition state theories in Gas/Condensed Phase, or more in general in Phase Space

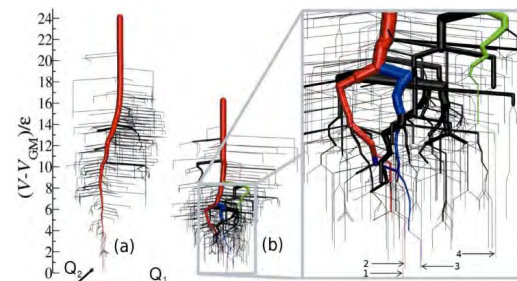
PNAS 98 7666 (2002) ; PRL 97, 028302 (2006); PRL 105,048304, (2010); PRL 106, 054101 (2011); PRL 115, 093003 (2015); *Nonlinearity* 28, 2677 (2015) ; *JCP (Perspective)* 155, 210901 (2021)

3 : Constructing energy landscape/network from Experimental data (data science)

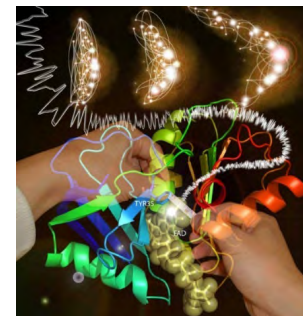
- ✓ Single Molecule Time series Analysis for Molecular Science

4 : Accelerate designing experiments via adaptive feedback between measurement/computation and data analysis

(reinforcement learning) Mach. Learn. 109, 327-372 (2020); *AISTATS*, Volume 206, 29pages (2023). PNAS under revision



PNAS 103, 18551 (2006)



PNAS 104,19297 (2007); PNAS 105, 536 (2008); PRL 111, 58301(2013); *Sci. Rep.* 5, 9174 (2015); *Nature Commun.* 6,10223 (2015); *JCP* 148, 123325 (2018) (special issue on single molecule biophysics)

Inferring cell-cell interactions from images

S. Sattari et al., *Sci. Adv.* 8(6), abj1720 (2022)
(Chosen as featured on-line article)

U. Basak et al, *PRE* 102, 012404 (2020); *JCP* 154, 034901 (2021)

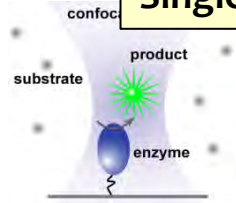


Fluorescence imaging

Facts and artifacts in interpreting dynamic disorder of enzymatic turnover reactions

T. Terentveva et al., *ACS Nano* 6 346-354(2012)

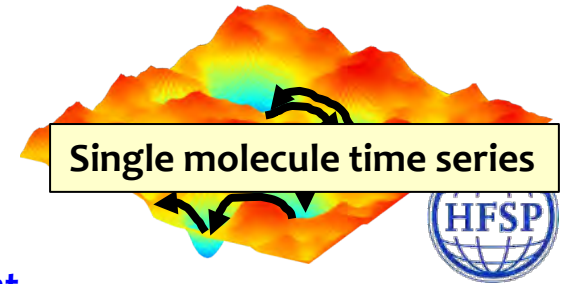
Single molecule time series



Single molecule time series

C.-B. Li et al., *PRL* 111, 58301(2013)

multidimensional Energy Landscape



Single molecule time series

A. Baba et al., *PNAS* 104,19297 (2007); *PCCP*, 13, 1395 (2011); K. Kamagata et al., *JACS* 134, 11525 (2012); J.N. Taylor et al., *Sci. Rep.* 5, 9174 (2015)
J. Alfermann et al., *Nature Chem. Bio.* 13, 1009 (2017)

Cell state defined by single cell

Raman Raman imaging

K. M. Helal, et al., *FEBS Lett.* 593, 2535-2544 (2019)(Chosen as cover art & highlight article)

J. Nicholas Taylor, et al. *J. Phys. Chem. B* 123, 4358-4372 (2019)

Let the Systems Speak for Themselves

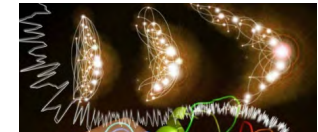
Coordination of microtubules beyond single cells in sepals



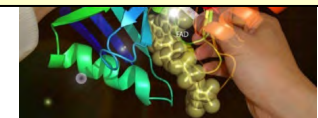
AFM

S. Tsugawa et al., *J. Phys. Chem. B* 110, 1836 (2016) (Chosen as highlighted article)
L. Hong et al., *Dev. Cell* 38, 15 (2016) (Chosen as cover art & Highlight)

Nonequilibrium steady state network



Single molecule time series



C.-B. Li et al., *PNAS*, 105, 536 (2008); C.-B. Li et al., *JPC B*, 113, 14732 (2009) ; T. Sultana et al. *JCP* 139, 245101(2013)

New change point analysis & m

Single molecule time series of F1 motor

C.-B. Li et al., *Nature Commun.* 6,10223 (2015)

Logr conc

Single molecule time series of F1 motor

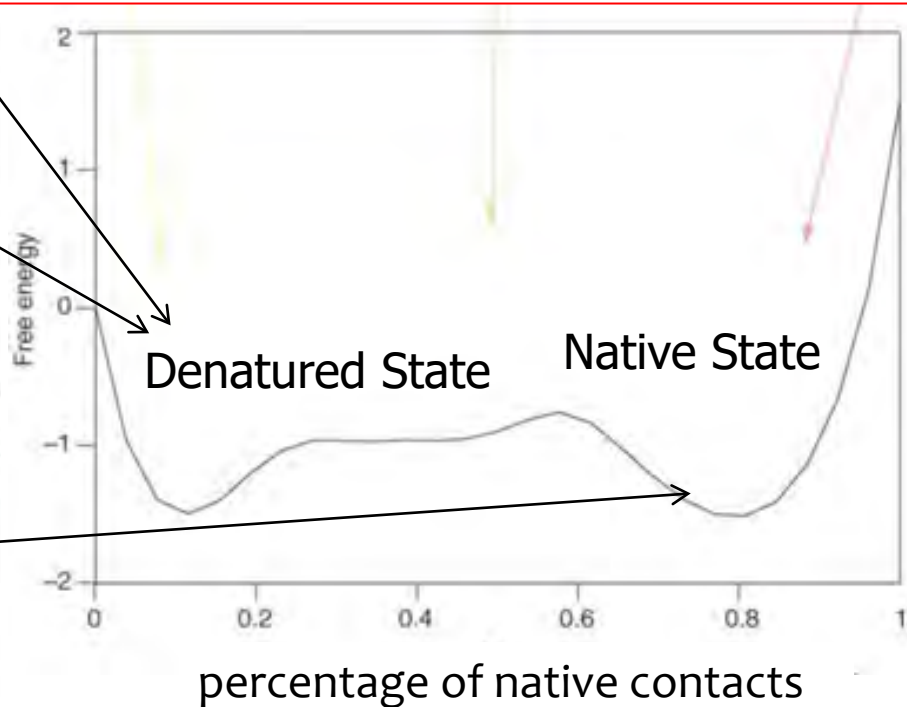
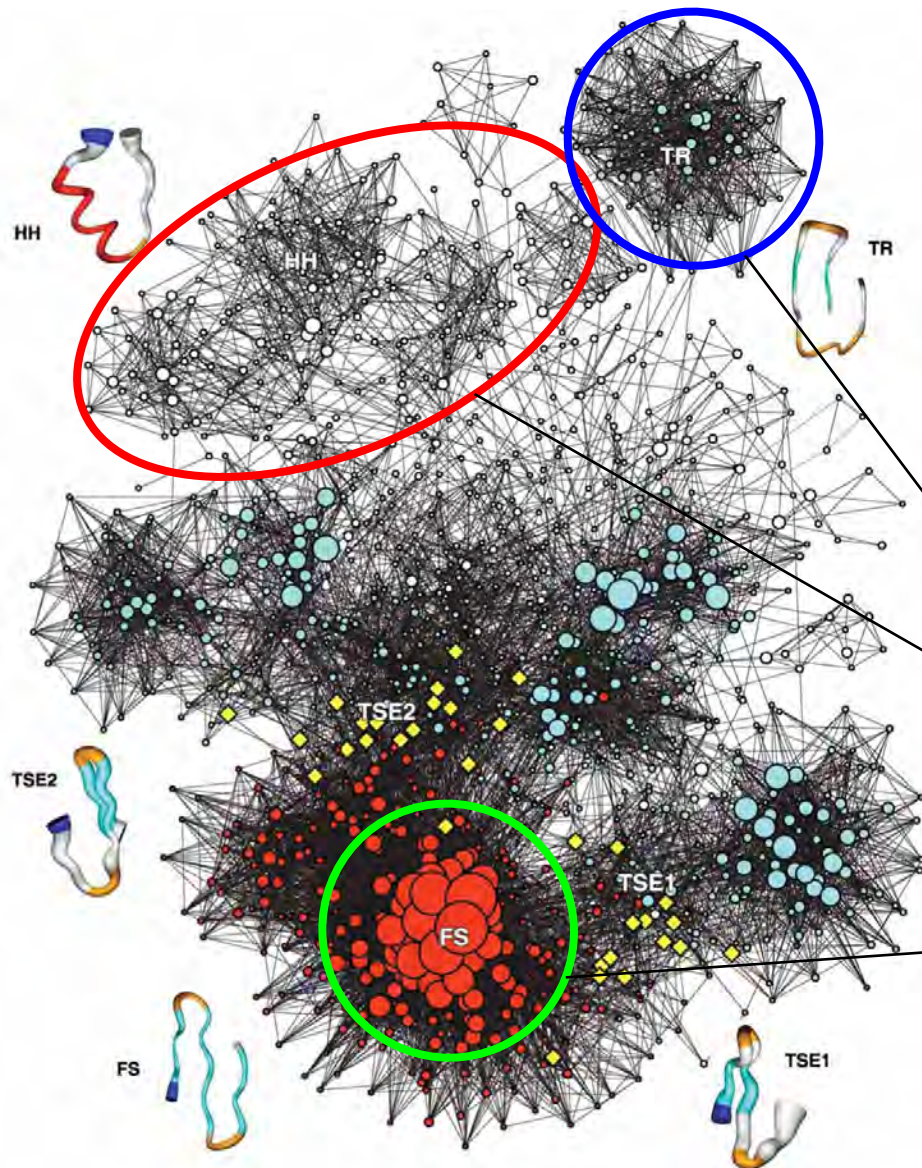
Yagin (Chosen as nigmgntea article)

Energy Landscape: conformation space network

F. Rao and A. Caflisch, *J. Mol. Biol.* (2004) **342**,299
Krivov, Muff, Caflisch, Karplus, *JPCB* **112**, 8701 (2008)

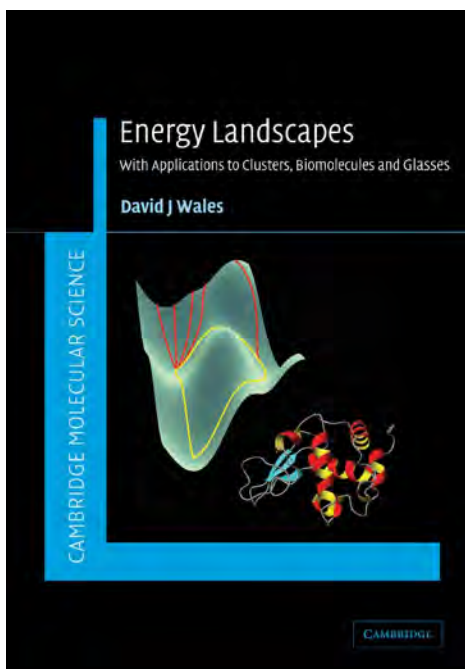
The complexity of conformation space network is masked by a projection

How can one elucidate multiple paths from an energy basin to one another?



Disconnectivity Graph: Visualization of Complexity of Energy Landscape

Disconnectivity Graph



D. J. Wales, *Energy Landscapes*
(Cambridge University Press, Cambridge, 2004).

Disconnectivity Graph

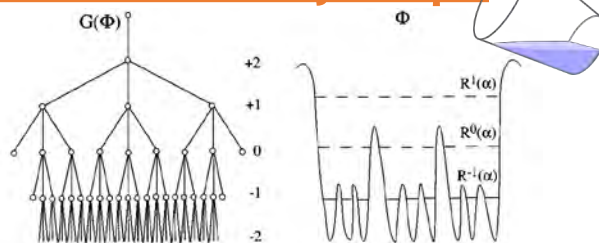


FIG. 2. A schematic one-dimensional partial cross section through a rough $3N$ -dimensional potential energy hypersurface Φ (triadic Koch curve). The horizontal lines on the PES Φ correspond to the energies used by the map $M^a(r;E)$ to define the $R(\alpha)$ basins. These energies determine the corresponding levels on the disconnectivity graph $G^E(\Phi)$.

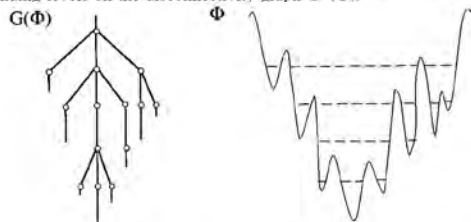


FIG. 4. A schematic representation of a funnel potential energy surface Φ and its corresponding $G^E(\Phi)$ graph.

O. M. Becker and M. Karplus,
J. Chem. Phys. **106**, 1495 (1997)

connectivity tree

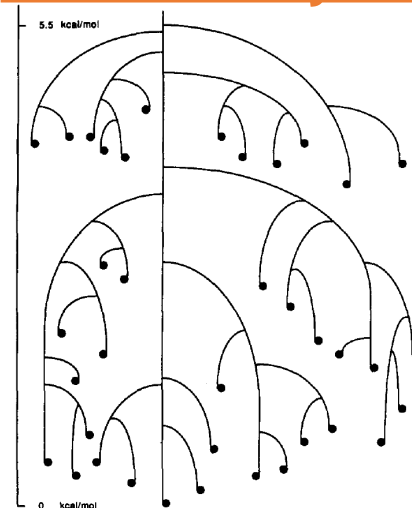


FIG. 9. A schematic representation of the connectivity tree (Cayley tree) of the minima and the transition states in the tetrapeptide IAN. Only the vertical direction which denotes the energy in kcal/mol is meaningful. The black circles (\bullet) are the minima and the tree junctions are the barriers. We emphasize that only the lowest energy path between the minima or clusters is shown and no conclusions on the topology of the connectivities should be drawn from the figure.

R. Czerminski and R. Elber,
J. Chem. Phys. **92**, 5580 (1990)

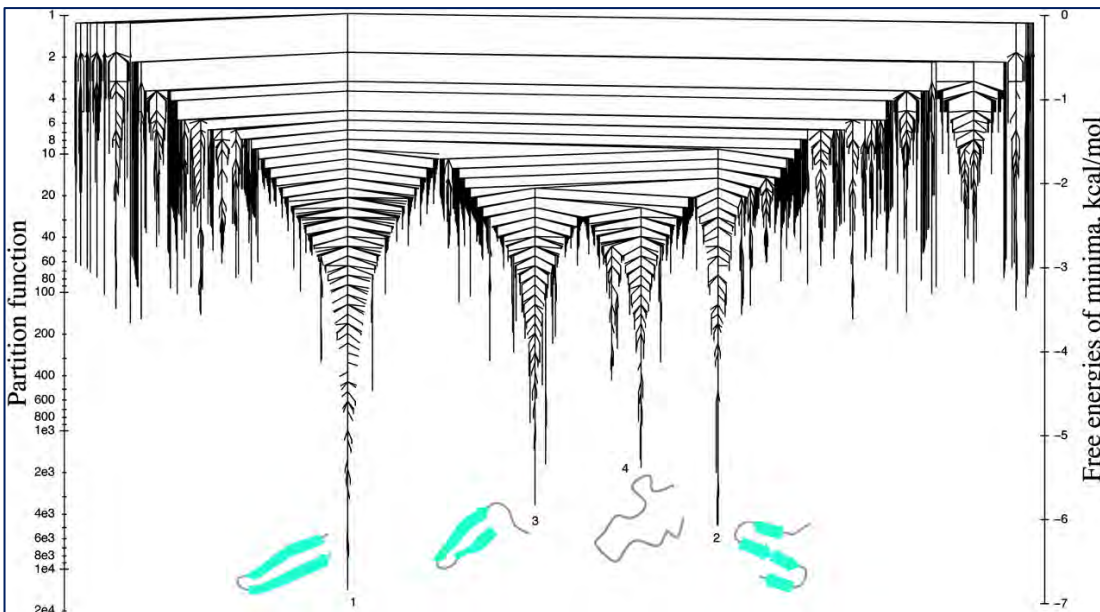
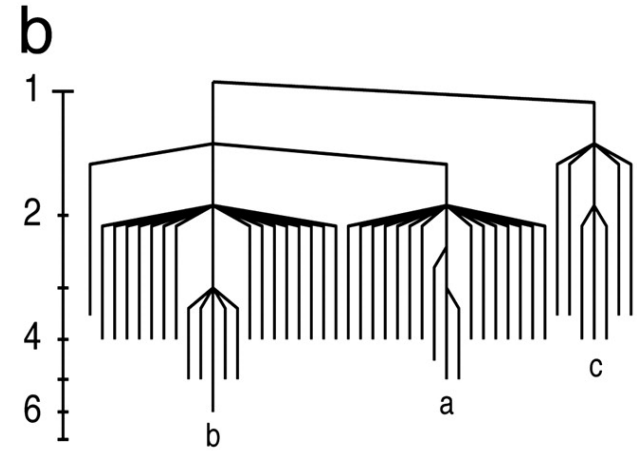
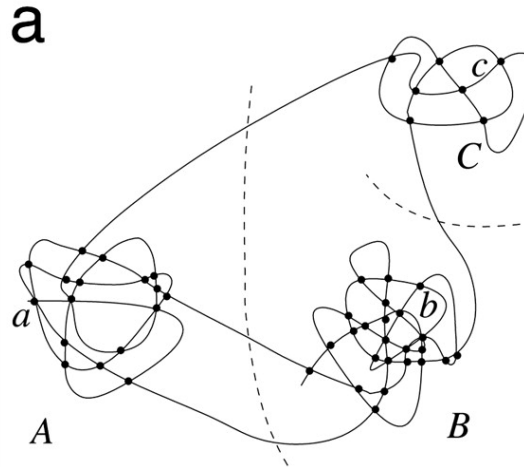
1. Background and Motivation
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Transition disconnectivity graph (TRDG)

S.V. Krivov & M. Karplus, Proc. Natl. Acad. Sci. USA **101** 14766 (2004)

Krivov, Muff, Caflisch, Karplus, JPCB **112**, 8701 (2008)

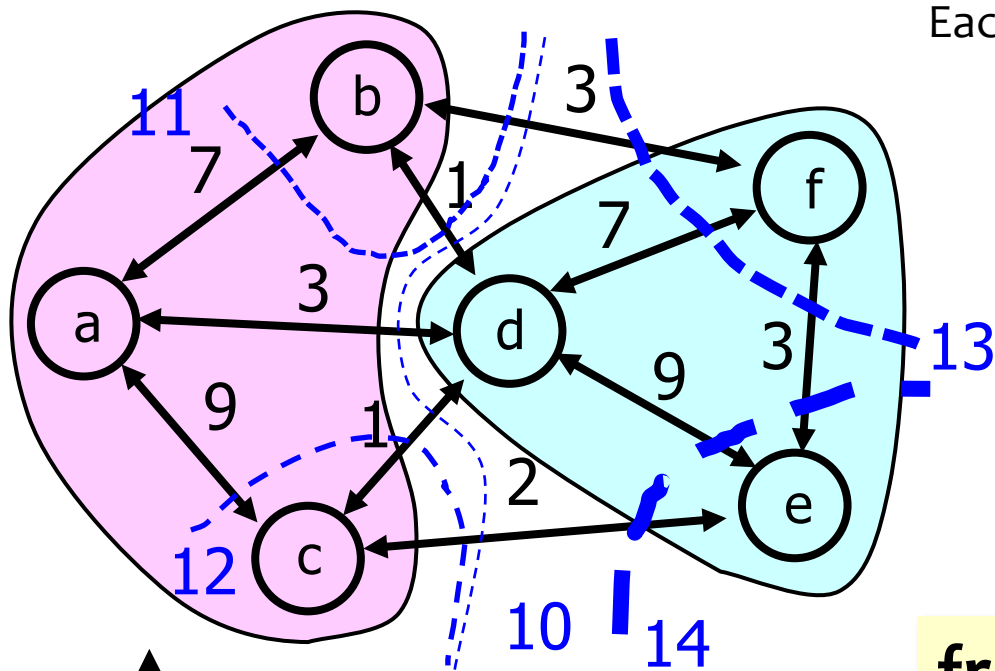
Transforming a conformation space network into a new type of **disconnectivity graph** in terms of a network theory using rates among conformations



TRDG of β -hairpin at 360 K.

Transformation from Network to Energy Landscape

S. V. Krivov and M. Karplus, *J. Chem. Phys.*, **117**, 10894 (2002); *Proc. Natl. Acad. Sci. USA* **101** 14766 (2004)

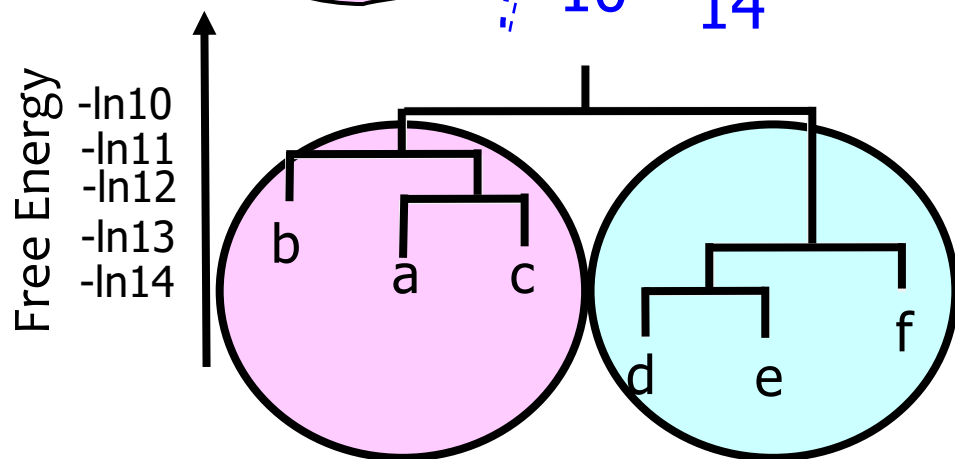


Each sub region has its own bottleneck.

$$\text{Energy Barrier} \propto -k_B T \ln(k_{i \rightarrow j} P_i)$$



Hierarchical grouping



free energy disconnectivity graph can be constructed by **taking into account all possible pathways in the network**

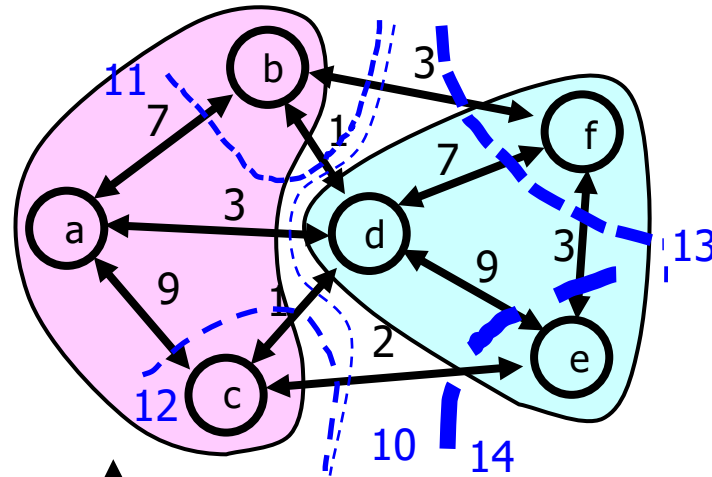
Free energy of the state i $F_i = -k_B T \ln \left(\frac{\sum_j N_{i \rightarrow j}}{N} \right) = -k_B T \ln \left(\frac{N_i}{N} \right) = -k_B T \ln P_i$

Kramers rate theory $k_{i \rightarrow j} \approx \frac{1}{\tau_0} \exp \left(-\frac{F_{ij} - F_i}{k_B T} \right)$ τ_0 : the decay timescale of the autocorrelation for motion exerted by friction from the environment.

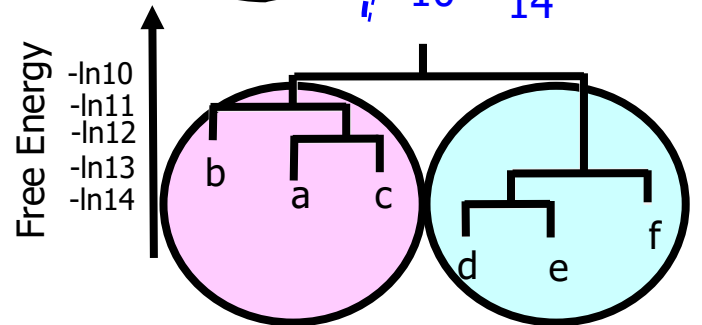
$k_{i \rightarrow j} P_i = \left(\frac{N_{i \rightarrow j}}{N_i} \frac{1}{\tau_{obs}} \right) \left(\frac{N_i}{N} \right) = \frac{1}{\tau_0} \exp \left(-\frac{F_{ij} - F_i}{k_B T} \right) \exp \left(-\frac{F_i}{k_B T} \right) = \frac{1}{\tau_0} \exp \left(-\frac{F_{ij}}{k_B T} \right)$.

Timescale of observation

$F_{ij} = -k_B T \ln(\tau_0 k_{i \rightarrow j} P_i) = -k_B T \ln \left(\frac{N_{i \rightarrow j}}{N} \frac{\tau_0}{\tau_{obs}} \right)$



in order to validate the free energy at the barrier, $F_{ij} = F_{ji}$
 $k_{i \rightarrow j} P_i = k_{j \rightarrow i} P_j$ should hold.



Why does the minimum flux approach miss the “transition state” in network?

When the space of optimization is the entire one, the minimum flux surface can be located near the “edge”

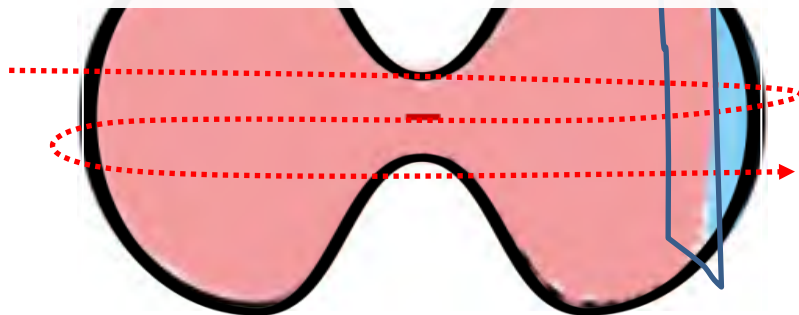
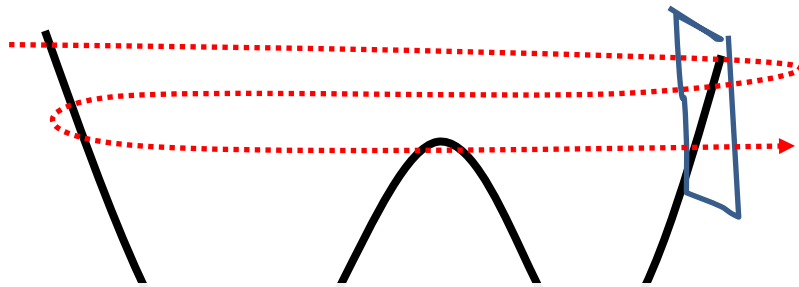
Can one infer the region of dividing surface over an

However, “Balanced min-cut is not unique and can break balance in some cases”

Nagahata, Wales, Komatsuzaki, *prepared for publication*

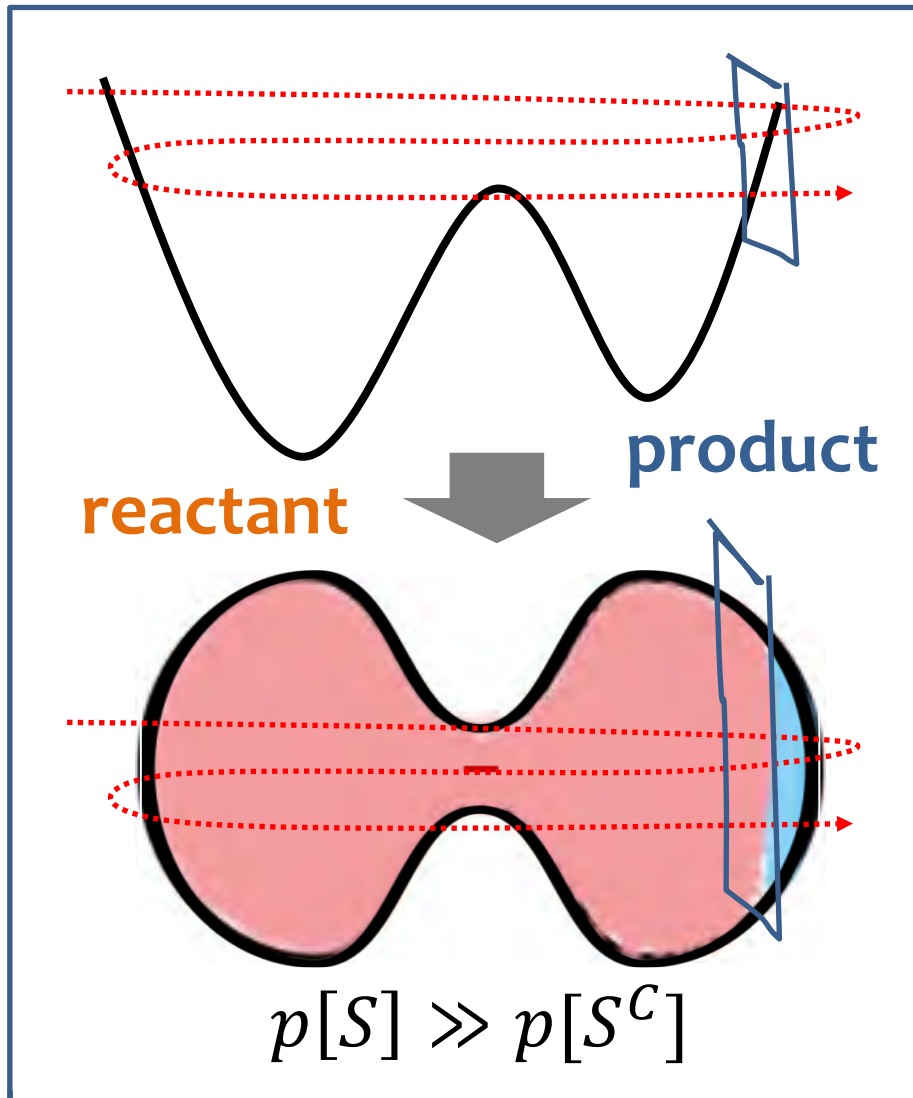
*Krivov-Karplus' choice

Balanced min-cut



$$p[S] \gg p[S^c]$$

Why does the minimum flux approach miss the “transition state” in network?



When the space of optimization is the entire one, the minimum flux surface can be located near the “edge”

Can one infer the region of dividing surface over an entire network?

*Krivov-Karplus' choice
Balanced min-cut

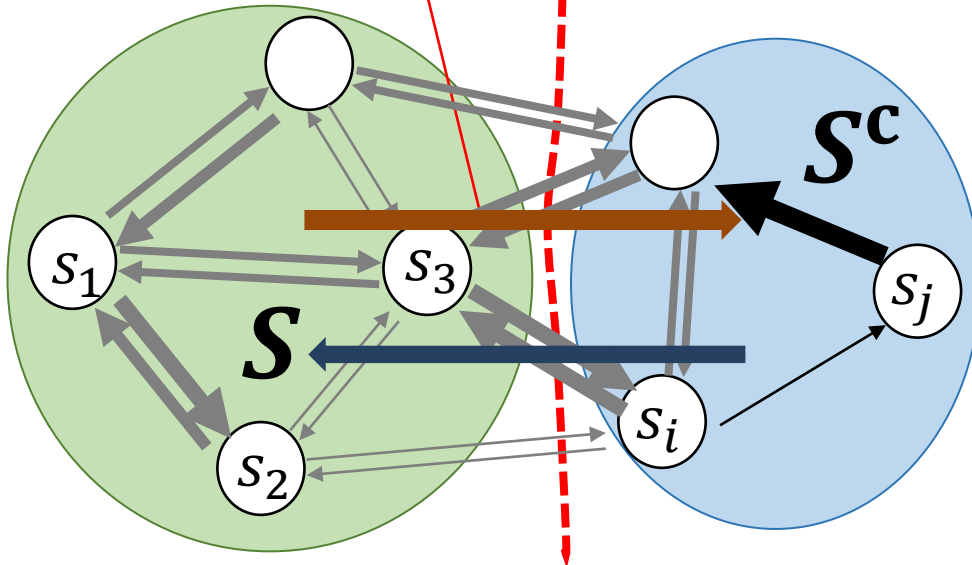
1. Background and Motivation
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TS in network is defined as a dividing surface across which passage is slowest



Y. Nagahata et al. *J. Phys. Chem. B* 120, 1961(2015)

$$k_{\text{network}}(S \rightarrow S^c) = p_{\Delta t}[S^c, S] / p[S]$$



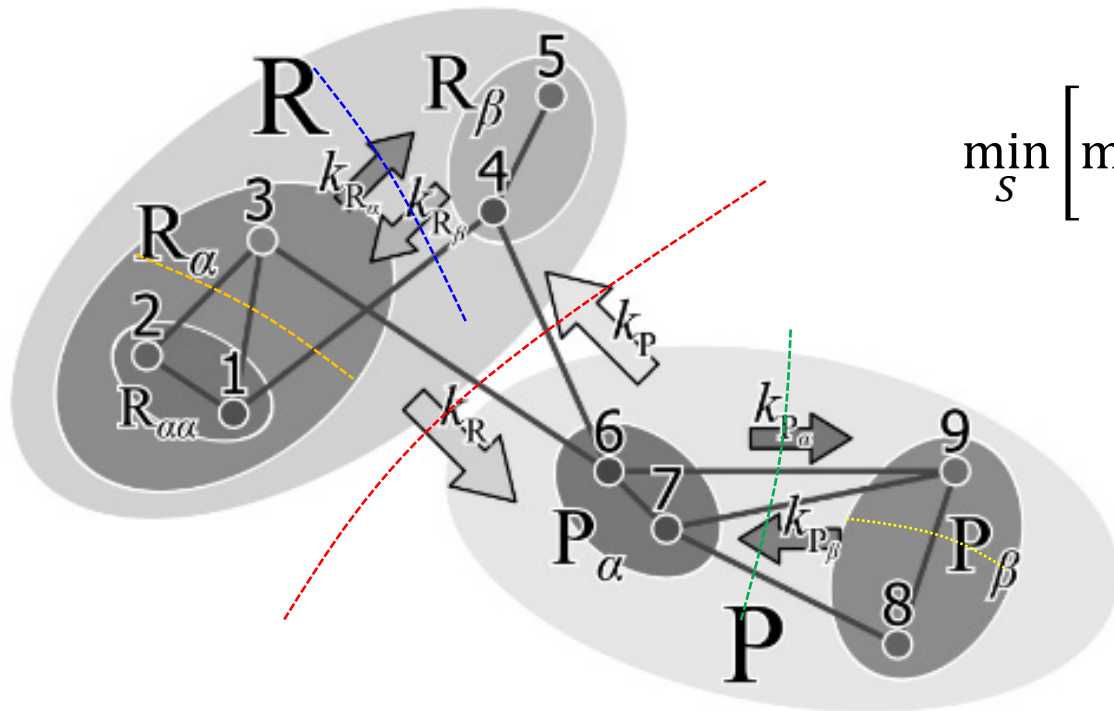
The quantity to be minimized is **NOT the flux** **BUT the gross rate constant** between the subsets of the network

Gross rate constants of $S \rightarrow S^c$ and $S^c \rightarrow S$ Gross rate (in time Δt) of $S \rightarrow S^c$ and $S^c \rightarrow S$

$$\min_S \left[\max \left\{ \frac{p_{\Delta t}[S^c, S]}{p[S]}, \frac{p_{\Delta t}[S, S^c]}{p[S^c]} \right\} \right]$$

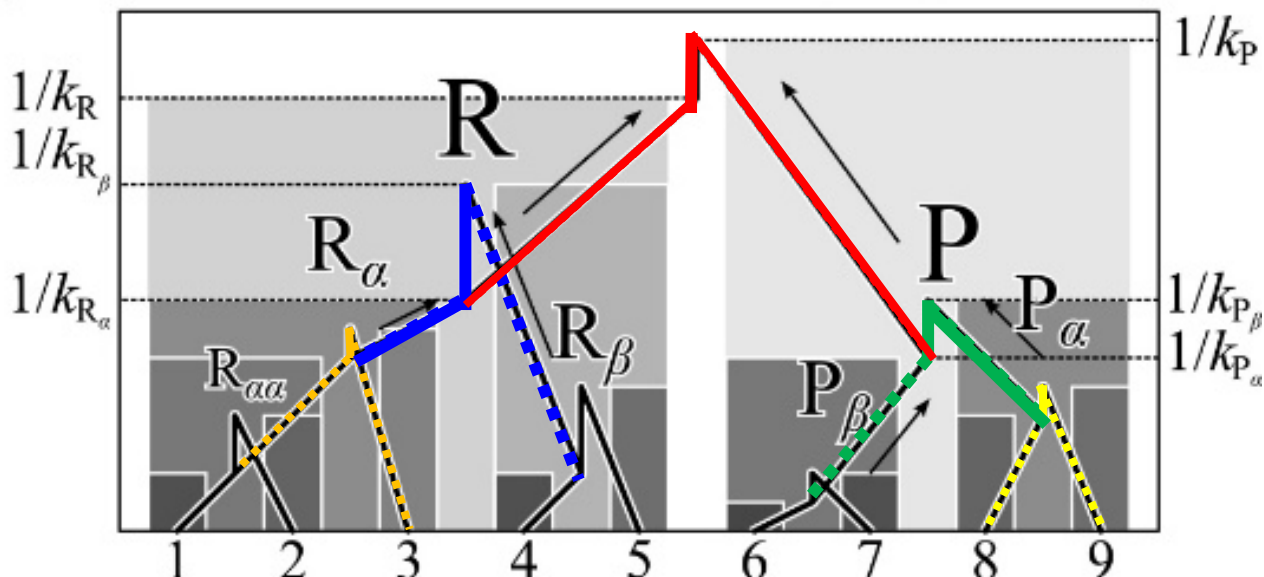
taking into account "the volume"

(a)

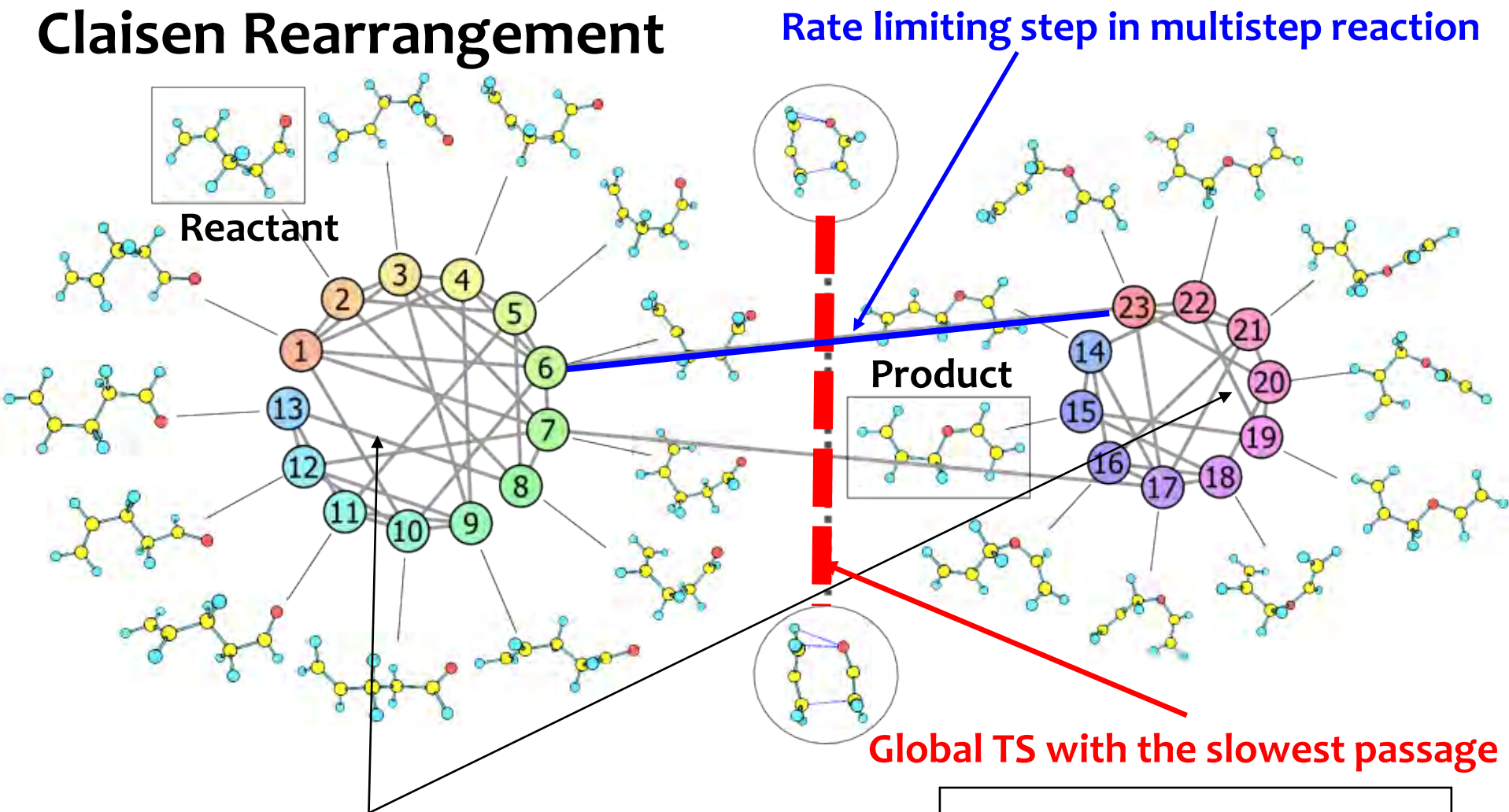


$$\min_S \left[\max \left\{ \frac{p_{\Delta t}[S^c, S]}{p[S]}, \frac{p_{\Delta t}[S, S^c]}{p[S^c]} \right\} \right]$$

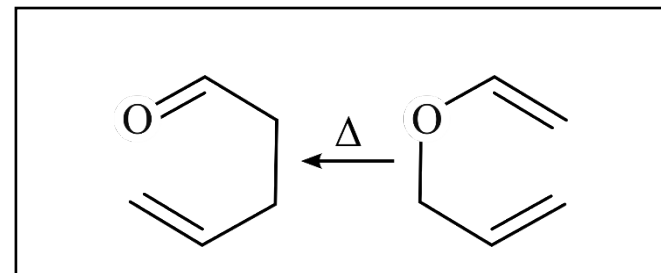
$$p[S] \left(= \sum_{s_i \in S} p_{eq}[s_i] \right)$$

(b) escaping timescale = $1/k$ ($\propto \ln[\Delta F^\ddagger]$)

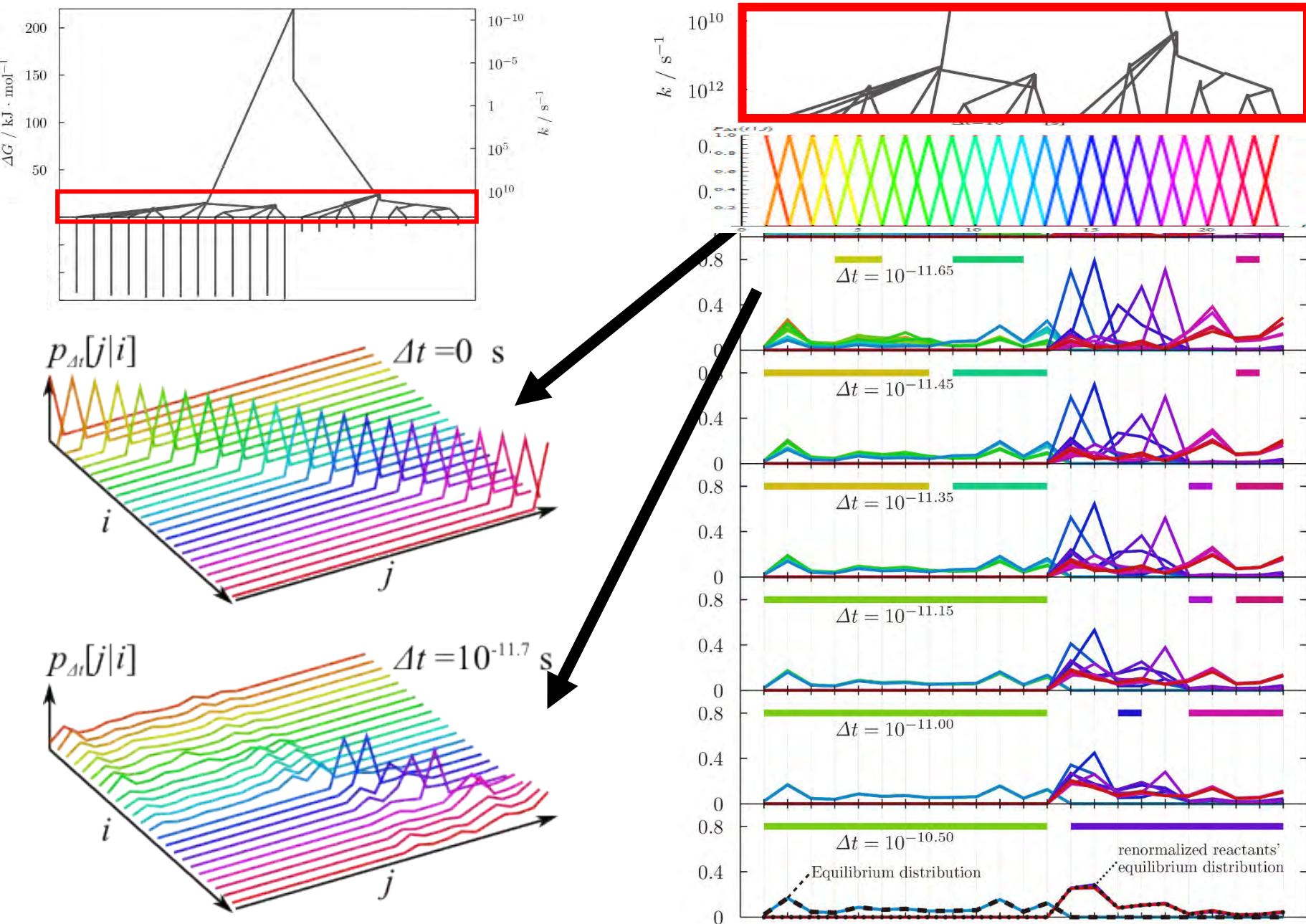
Reaction Network of Claisen Rearrangement

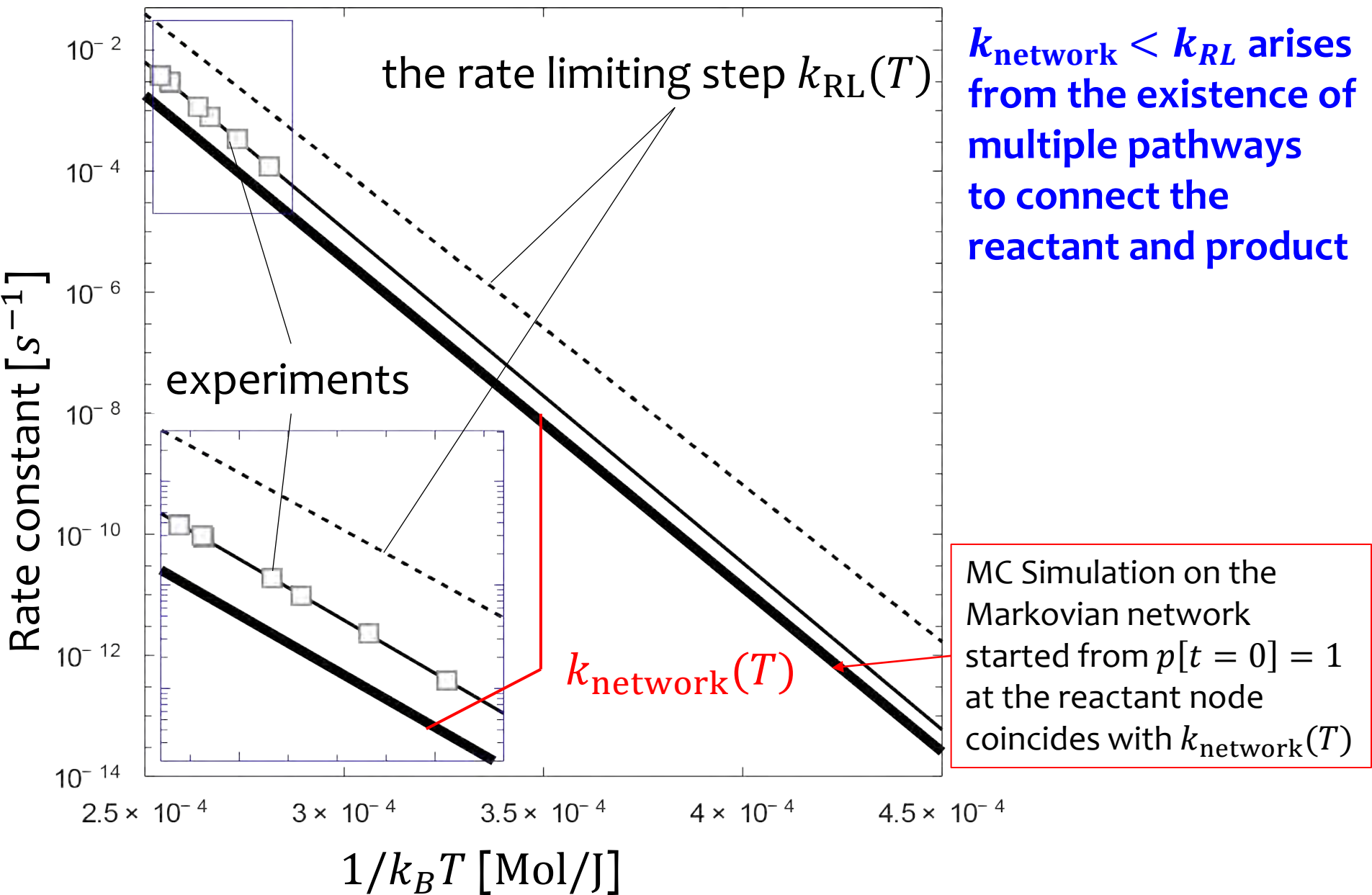


Each edge between nodes has a saddle linking the two minima & an estimated TST rate constant at 473.15 K



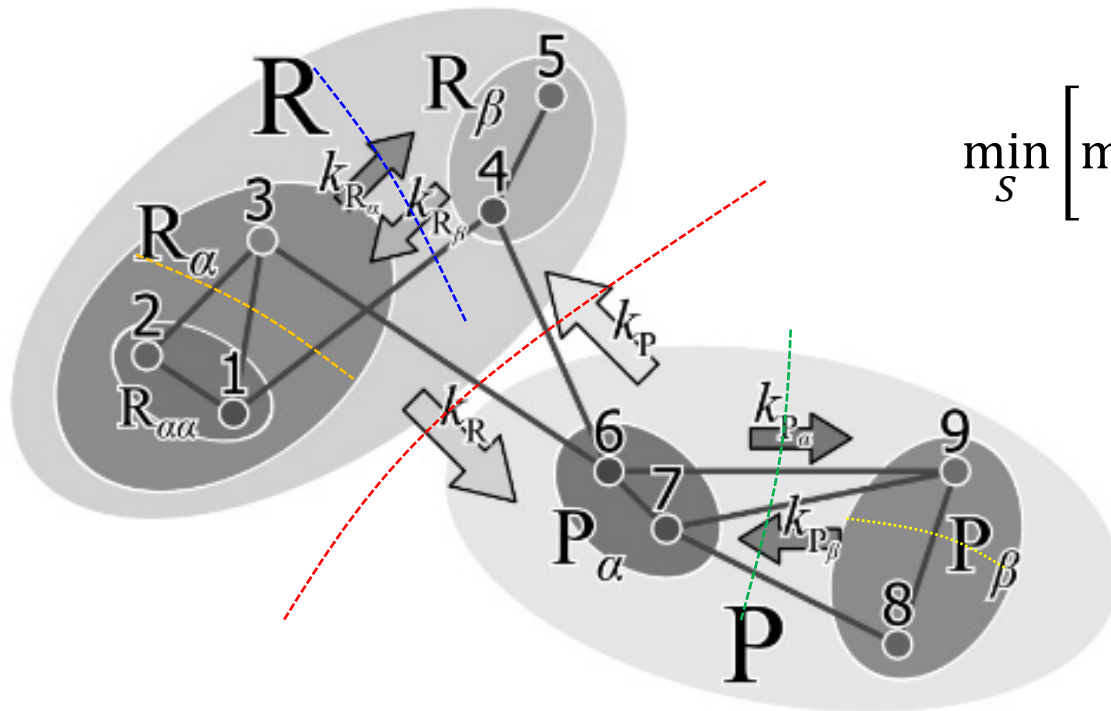
Kinetic disconnectivity graph and Markov kinetics





Experimental data F. W. Schuler and G. W. Murphy, J. Am. Chem. Soc. 72, 3155 (1950).

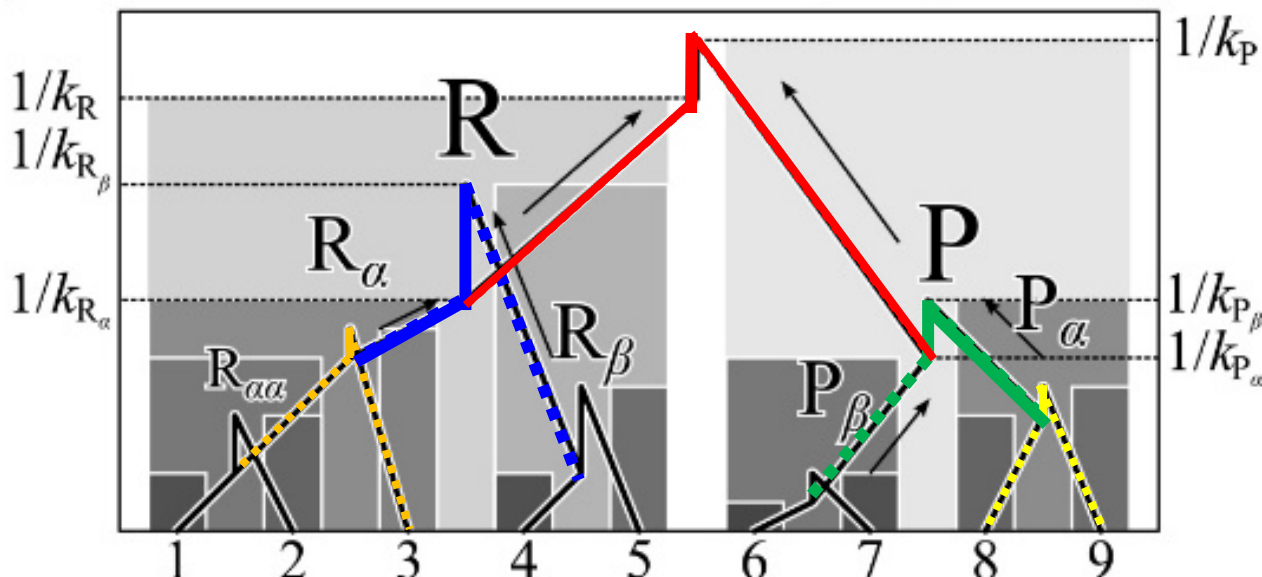
(a)



$$\min_S \left[\max \left\{ \frac{p_{\Delta t}[S^c, S]}{p[S]}, \frac{p_{\Delta t}[S, S^c]}{p[S^c]} \right\} \right]$$

$$p[S] \left(= \sum_{s_i \in S} p_{eq}[s_i] \right)$$

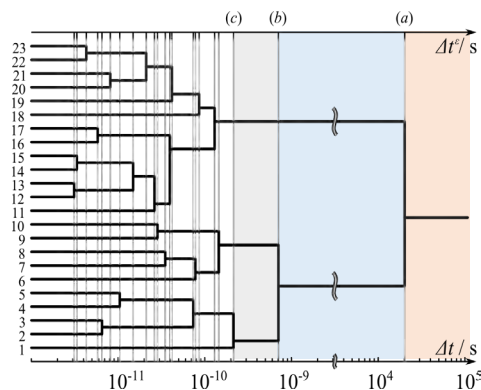
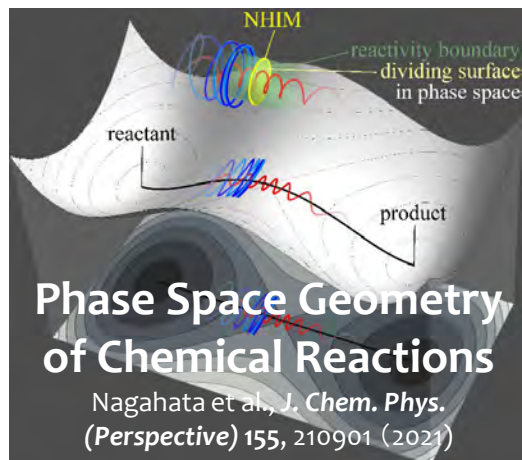
(b) escaping timescale = $1/k$ ($\propto \ln[\Delta F^\ddagger]$)



This requires the assumption of thermal equilibrium irrespective of the pair of S, S^c

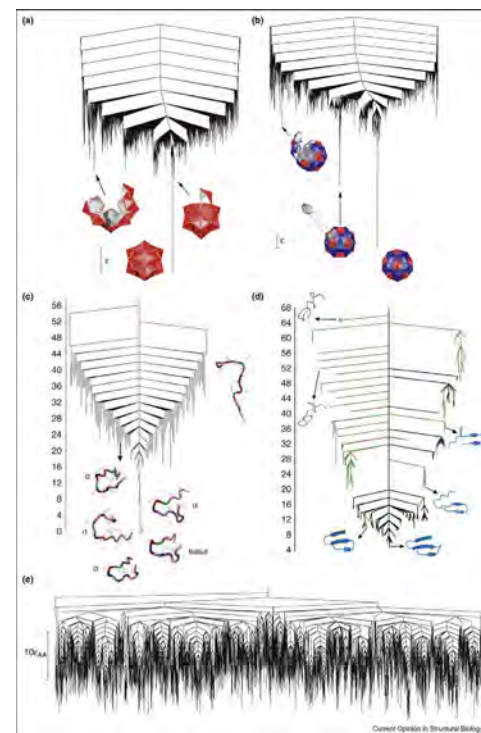
1. Background and Motivation
2. Transition Disconnectivity Graph
3. Network Transition States
- 4. Lumping Theory to Preserve Hierarchical Timescales**
5. Outlooks

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Representation of Kinetic Hierarchies not necessarily based on equilibrium

Free from stationarity, timescale separation, Predefining subnetworks



D. J. Wales, *Curr. Opin. Struct. Biol.* 20, 3 (2010).

Landscape

Dynamics

Lumpability diagram vs kinetic DG

Lumpability diagram

Identifying ϵ -indistinguishability of dynamics under a given timescale of observation.

Computational costs:

Solving ODE:

diagonalization $\mathcal{O}(n^3)$

Find Δt^ϵ :

matrix multiplication $\mathcal{O}(n^3)$

matrix arithmetic $\mathcal{O}(n^2)$

multi-precision

root finding $\times \mathcal{O}(n^2)$

Constructing diagram

complete linkage $\mathcal{O}(n^2)$



Dr. Yutaka Nagahata



Kinetic disconnectivity graph

Showing whether each subnetwork is connected under a given threshold of timescale

Computational costs:

Computing rate constants:

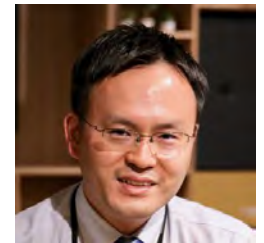
arithmetic $\times \mathcal{O}(n^2)$

Constructing diagram

single linkage $\mathcal{O}(n^2)$



Prof. Tetsuya Taketsugu



Prof. Satoshi Maeda