Estimation of low-energy refolding paths
Visualization of Lattice Protein Dynamics

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Outline

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The HP-model

Suggested by Dill, Chan and Lau in the late 1980ies. In this *simplified model*, a conformation is a *self-avoiding walk (SAW)* on a given lattice in 2 or 3 dimensions. Each bond is a straight line, bond angles have a few discrete values. The 20 letter alphabet of amino acids (monomers) is reduced to a two letter alphabet, namely **H** and **P**. H represents hydrophobic monomers, P represents hydrophilic or *polar* monomers.

Advantages:

- lattice-independent folding algorithms
- simple energy function
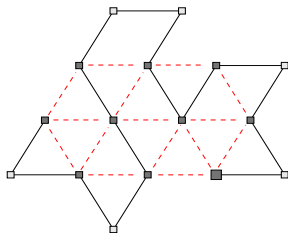- hydrophobicity can be reasonably modeled
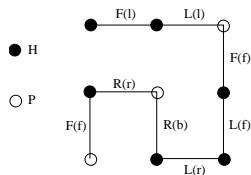
FRRLLFLF
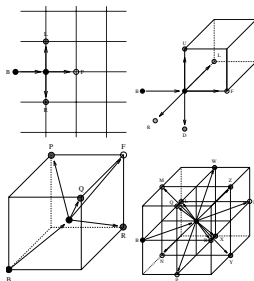
Contact Potentials

Generally, the energy function for a sequence with n residues $\mathfrak{S} = \mathfrak{s}_1\mathfrak{s}_2\ldots\mathfrak{s}_n$ with $\mathfrak{s}_i \in \mathscr{A} = \{a_1, a_2, \ldots, a_b\}$, the alphabet of b residues, and an overall configuration $x = (\mathbf{x}_1, \mathbf{x}_2, \ldots, \mathbf{x}_n)$ on a lattice $\mathscr{L}$ can be written as the sum of pair potentials

$$E(\mathfrak{S}, x) = \sum_{\substack{i < j-1 \\ |\mathbf{x}_i - \mathbf{x}_j| = 1}} \Psi[\mathfrak{s}_i, \mathfrak{s}_j].$$
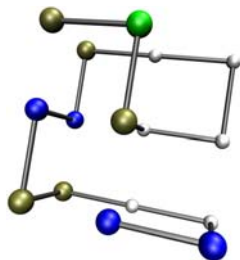
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Lattice proteins

$\mathfrak{S} = $ HPHPHHHPPHHHHPHPH $n = 16$

$E = -15$

	H	P
H	−1	0
P	0	0

$\mathfrak{S} = $ NNHHPPNNPHHHHPXP $n = 16$

$E = -16$

	H	P	N	X
H	−4	0	0	0
P	0	1	−1	0
N	0	−1	1	0
X	0	0	0	0

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Lattice proteins - interaction scheme II

$\mathfrak{S} = $ HHHHNNNNHHHHHHHNHNHPNNNNNNNNPNPNNHNNHNHHHXXHHPXHNHHNXNHHNPHPNHHNHHNPXNHHHHHH

$$n = 74$$

$$E = -120$$
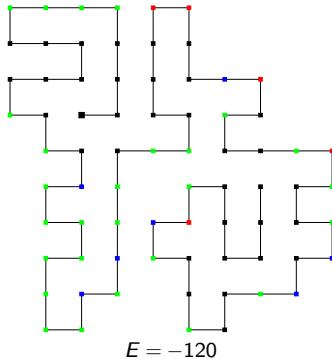
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H	−4	0	0	0
P	0	0	−1	0
N	0	−1	0	0
X	0	0	0	0

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Folding landscape - energy landscape

The energy landscape of a biopolymer molecule is a complex surface of the (free) energy versus the conformational degrees of freedom.

Number of lattice protein structures

$$c_n \sim \mu^n \cdot n^{\gamma-1}$$

problem is NP-hard

In the RNA case
$$c_n \sim 1.86^n \cdot n^{-\frac{3}{2}}$$

dynamic programming algorithms available

dim	Lattice Type	μ	γ
2	SQ	2.63820	1.34275
	TRI	4.15076	1.343
	HEX	1.84777	1.345
3	SC	4.68391	1.161
	BCC	6.53036	1.161
	FCC	10.0364	1.162

Formally, three things are needed to construct an energy landscape:

- A set X of configurations
- a symmetric neighborhood relation $\mathfrak{N} : X \times X$
- an energy function $f : X \to \mathbf{R}$

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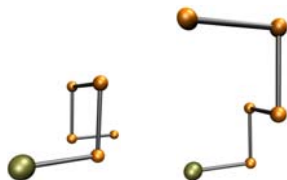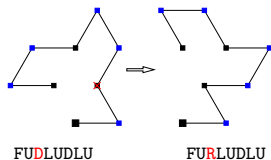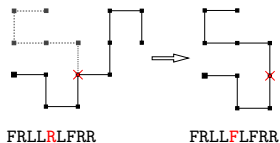$$c_n \sim 1.86^n \cdot n^{-\frac{3}{2}}$$

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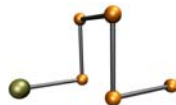Formally, three things are needed to construct an energy landscape:

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The move set
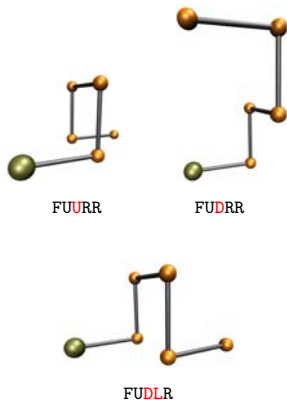

FRLL**R**LFRR $\Rightarrow$ FRLL**F**LFRR

FU**U**RR   FU**D**RR

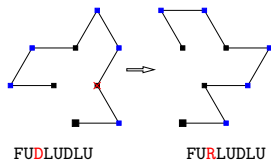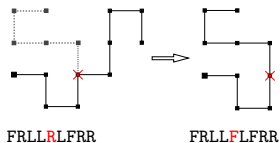FU**D**LUDLU $\Rightarrow$ FU**R**LUDLU

FU**D**LR

- For each move there must be an inverse move
- Resulting structure must be in X
- Move set must be *ergodic*

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The move set

FRLL**R**LFRR → FRLL**F**LFRR

FU**U**RR   FU**D**RR

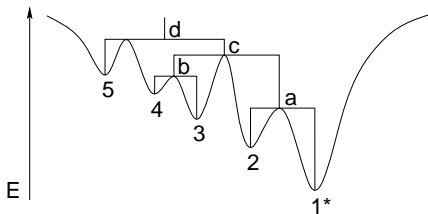FU**D**LUDLU → FU**R**LUDLU

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Energy barriers and barrier trees

Some topological definitions:
A structure is a

- **local minimum** if its energy is lower than the energy of **all** neighbors

- local maximum if its energy is higher than the energy of **all** neighbors

- saddle point if there are at least two local minima thar can be reached by a downhill walk starting at this point

We further define

- a walk between two conformations x and y as a list of conformations $x = x_1 \ldots x_{m+1} = y$ such that $\forall 1 \leq i \leq m : \mathfrak{N}(x_i, x_{i+1})$

- the lower part of the energy landscape (written as $X^{\leq \eta}$) as *all* conformations x such that $E(\mathfrak{S}, x) \leq \eta$ (with a predefined threshold η).

C. Flamm, I. L. Hofacker, P. F. Stadler, and M. T. Wolfinger.
Barrier trees of degenerate landscapes.
Z. Phys. Chem., 216:155–173, 2002.

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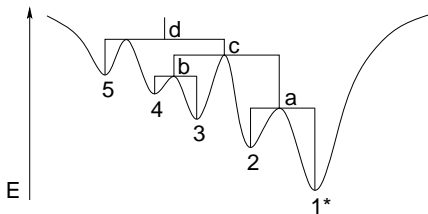
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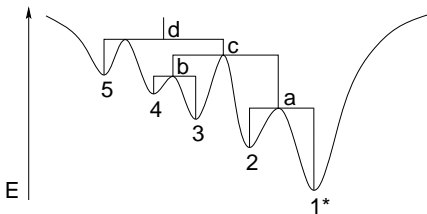
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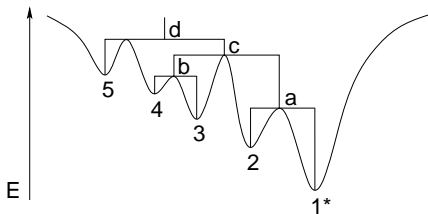
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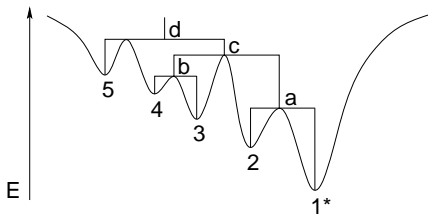
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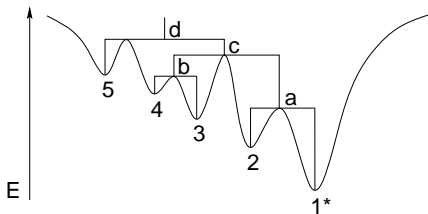
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Information from the barrier tree

- Local minima
- Saddle points
- Barrier heights
- Gradient basins
- Partition functions and free energies of (gradient) basins

This information can be used to approximate the dynamics of biopolymers, i.e. transition rates between different macrostates (basins in the barrier tree)

- $r_{\beta\alpha} = \Gamma_{\beta\alpha} \exp\left(-(E^*_{\beta\alpha} - G_\alpha)/kT\right)$

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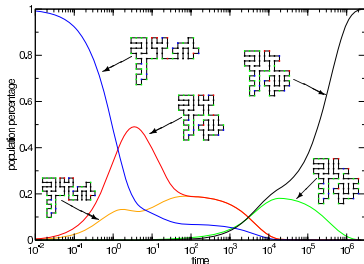
The lower part of the energy landscape

Two conformations x and y are mutually accessible at the level η (written as $x \leftarrow\!\!\!\!\multimap \underline{\eta} \multimap\!\!\!\!\rightarrow y$) if there is a walk from x to y such that all conformations z in the walk satisfy $E(\mathfrak{S}, z) \leq \eta$. The saddle height $\hat{f}(x, y)$ of x and y is defined by

$$\hat{f}(x, y) = \min\{\eta \mid x \leftarrow\!\!\!\!\multimap \underline{\eta} \multimap\!\!\!\!\rightarrow y\}$$

Given the set of all local minima $X_{\min}^{\leq\eta}$ below threshold η, the lower energy part $X^{\leq\eta}$ of the energy landscape can alternatively be written as
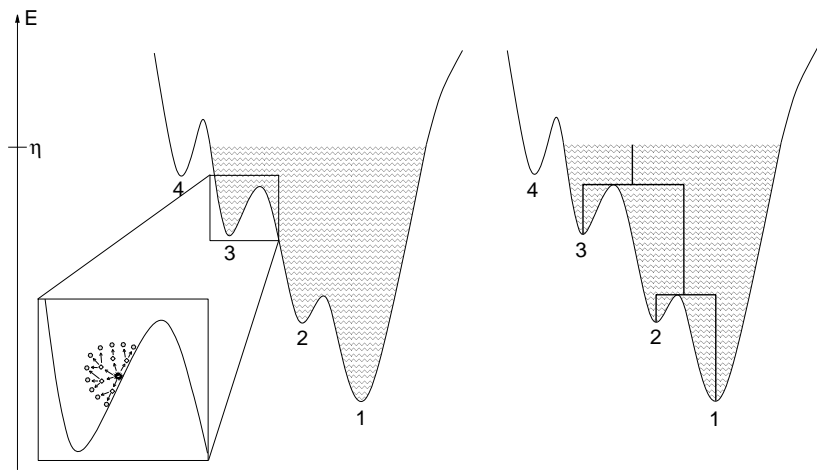
$$X^{\leq\eta} = \{y \mid \exists x \in X_{\min}^{\leq\eta} : \hat{f}(x, y) \leq \eta\}$$

Given a restricted set of low-energy conformations, X_{init}, and a reasonable value for η, the lower part of the energy landscape can be calculated.
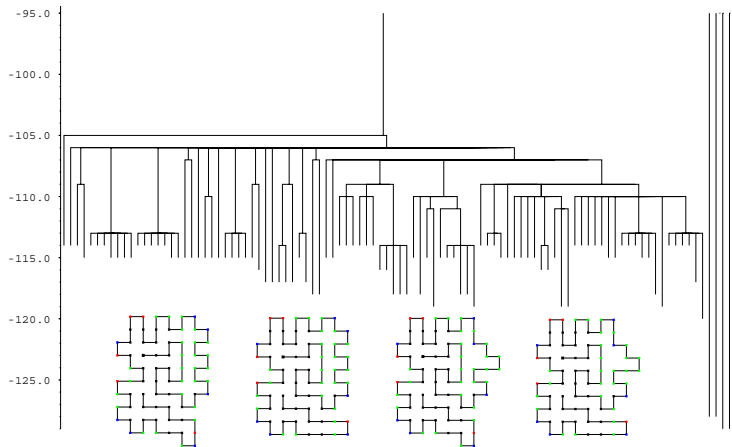
M. T. Wolfinger, S. Will, I. L. Hofacker, R. Backofen, and P. F. Stadler.
Exploring the lower part of discrete polymer model energy landscapes.
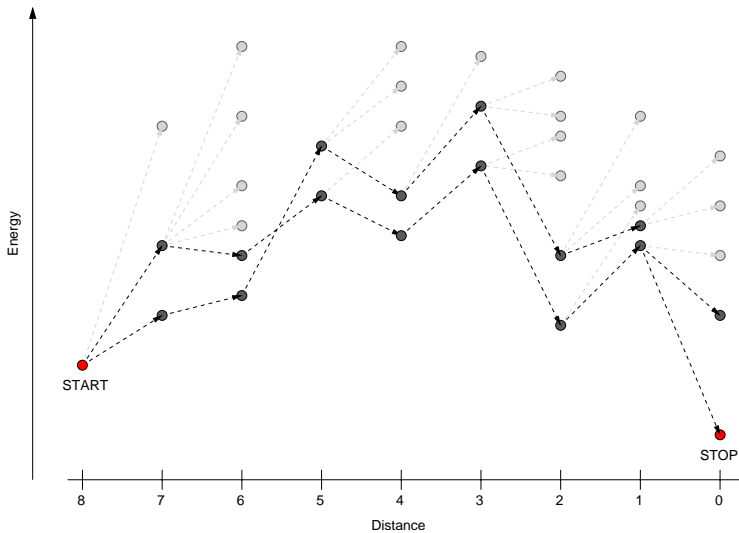Europhys. Lett., 2006.

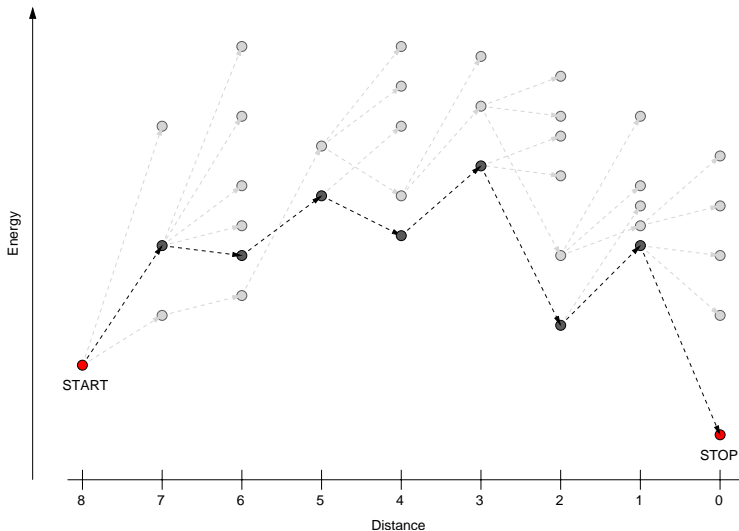*tbi*

The Flooder approach

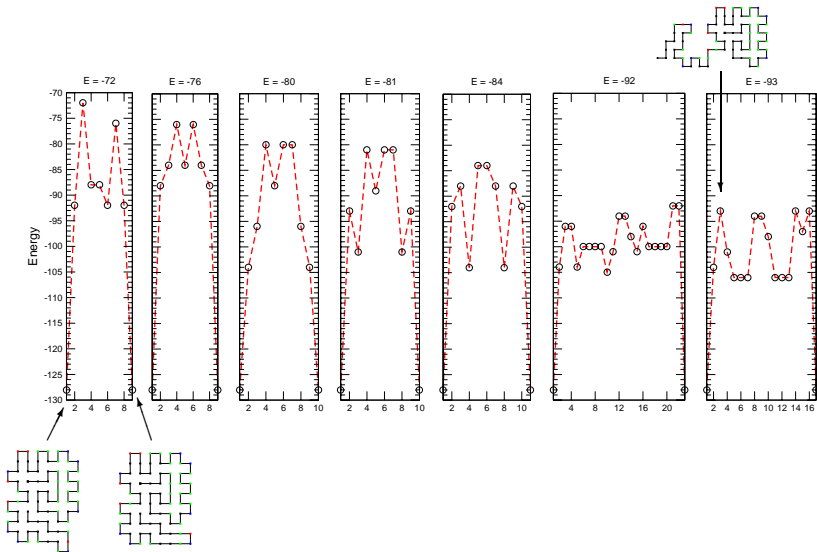# PathFinder - illustration

START

STOP

Energy

Distance

8 7 6 5 4 3 2 1 0

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PathFinder - illustration

START

STOP

Energy

Distance

8 7 6 5 4 3 2 1 0

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Refolding profiles

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The probability distribution P of structures as a function of time is ruled by a set of forward equations, also known as the master equation

$$\frac{dP_t(x)}{dt} = \sum_{y \neq x} [P_t(y) k_{xy} - P_t(x) k_{yx}]$$

Given an initial population distribution, how does the system evolve in time? (What is the population distribution after n time-steps?)

$$\frac{d}{dt} P_t = \mathbf{U} P_t \implies P_t = e^{t\mathbf{U}} P_0$$

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Barrier tree kinetics

For a reduced description we need

- macro-states that form a partition of full configuration space
- transition rates between macro-states, e.g.

$r_{\beta\alpha} = \Gamma_{\beta\alpha} \exp\left(-(E^*_{\beta\alpha} - G_\alpha)/kT\right)$ or

$r_{\beta\alpha} = \sum_{y\in\beta} \sum_{x\in\alpha} r_{yx} \mathrm{Prob}[x|\alpha] \quad \text{for } \alpha \neq \beta \text{ with } r_{yx} = \begin{cases} e^{\frac{-\Delta E}{kT}} & if\,\Delta E > 0 \\ 0 & y \notin \mathcal{N}(x) \\ 1 \end{cases}$

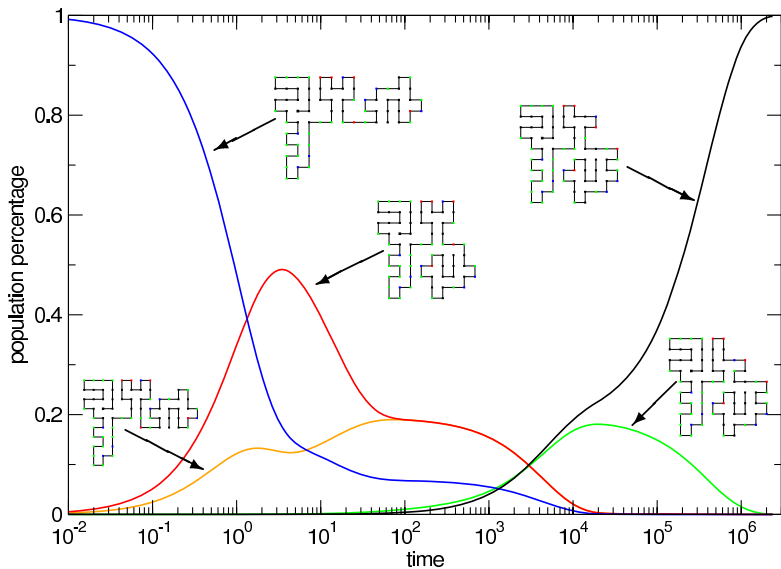All relevant quantities can be computed via the flooding algorithm.

M. T. Wolfinger, W. A. Svrcek-Seiler, C. Flamm, I. L. Hofacker, and P. F. Stadler.
Efficient computation of RNA folding dynamics.
J. Phys. A: Math. Gen., 37(17):4731–4741, 2004.

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Barrier tree kinetics - example

Kinetic Folding Algorithm

Simulate folding kinetics by a rejection-less Monte-Carlo type algorithm:

Generate all neighbors using the move-set

Assign rates to each move, e.g.

$$P_i = \min\left\{1, \exp\left(-\frac{\Delta E}{kT}\right)\right\}$$

Select a move with probability proportional to its rate
Advance clock $1/\sum_i P_i$.
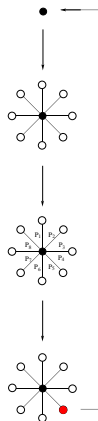
C. Flamm, W. Fontana, I. Hofacker, and P. Schuster.
RNA folding kinetics at elementary step resolution.
RNA, 6:325–338, 2000.

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Visualization

- Visualize Pinfold output
- Support data analysis
- Emphasize possible relationships
- Provide simulation comparison
- Provide data analysis and exploration
- Uncover regularities
- Expose the unseen
- Speed up coginition

Shneiderman's mantra:

"Overview first, zoom and filter, details on demand "

- Start with an overview
- Let the user filter out interesting data
- Show details only on demand for different data

S. Pötzsch, G. Scheuermann, M. T. Wolfinger, C. Flamm, and P. F. Stadler.
Visualization of lattice-based protein folding simulations.
In *10th International Conference on Information Visualization (IV06)*, 2006.

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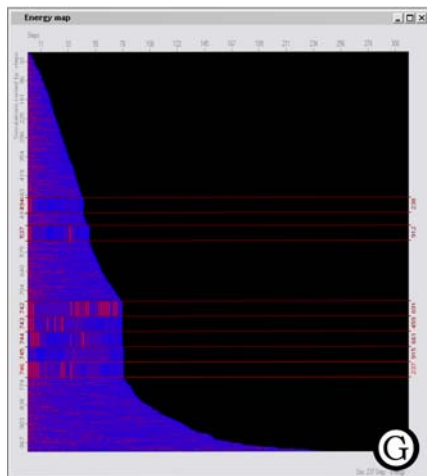"Overview first, zoom and filter, details on demand"

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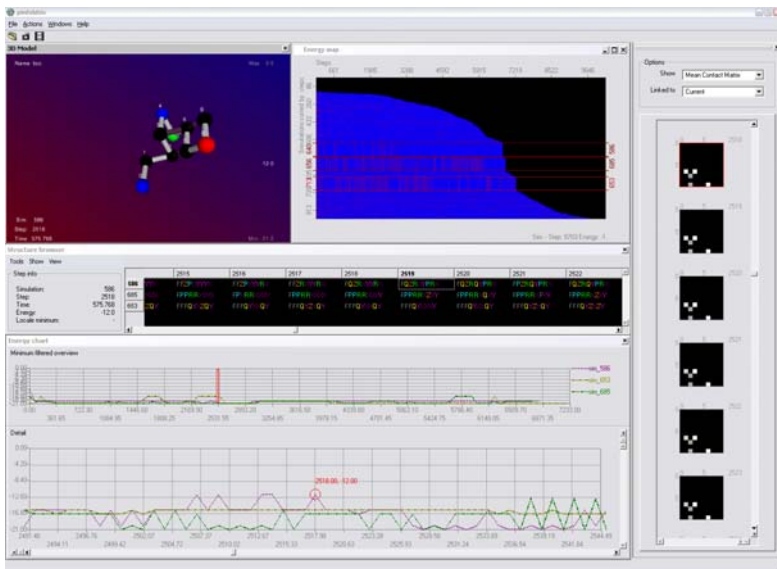
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Overview first

- Energy map
- Focus & context technique
- Huge data sets, limited screen size
- Details and overview in one window

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Details on demand

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Conclusion

- **Barrier trees** approximate the landscape topology and folding kinetics.

- A heuristic approach allows to sample low-energy refolding paths between different structures

- A macrostate approach of folding kinetics reduces simulation time drastically.

- A tool for visualization of folding trajectories enables a thorough investigation of folding kinetics simulations.

- This newly generated framework provides a powerful method for further refinement of biopolymer folding landscapes.

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Thanks

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Rolf Backofen

Peter Stadler

Peter Schuster

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