

Energy Landscapes of Biopolymers

Michael Wolfinger

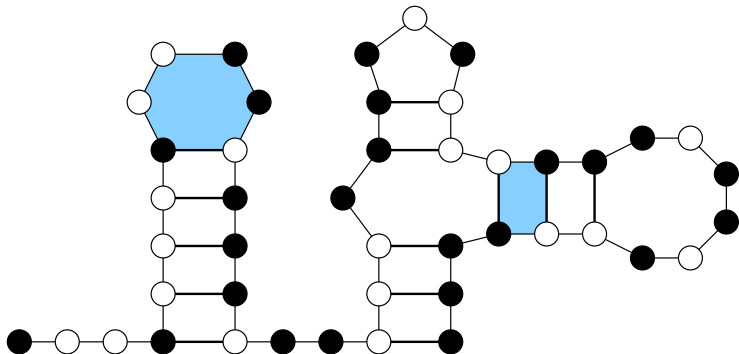
Institute for Theoretical Chemistry
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Outline

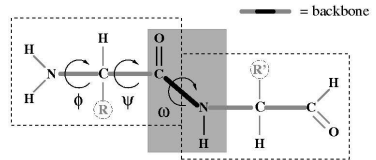
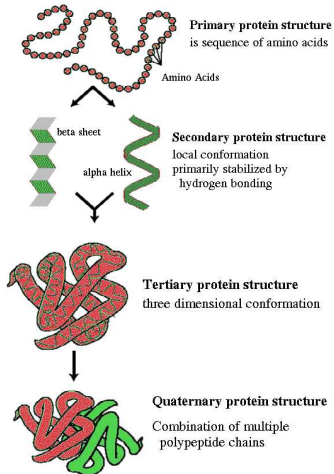
- 1 Biopolymers
- 2 Conformation space
- 3 Energy landscapes
- 4 Dynamics of biopolymers
- 5 Examples

RNA energy model



The energy of a sequence and particular structure is given as the sum of contributions from the "loops" (planar faces). Stacks yield stabilizing contributions, all other loops lead to destabilizing energy contributions.

Levels of structure in proteins

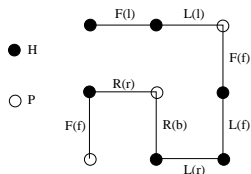


The HP-model

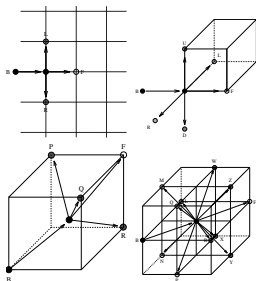
Suggested by Dill, Chan and Lau in the late 1980ies. In this *simplified model*, a conformation is a *self-avoiding walk (SAW)* on a given lattice in 2 or 3 dimensions. Each bond is a straight line, bond angles have a few discrete values. The 20 letter alphabet of amino acids (monomers) is reduced to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

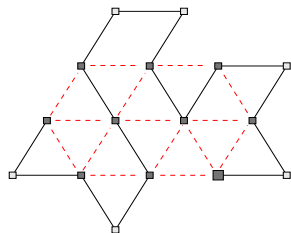


FRRLFLF



Lattice proteins

HPHPHHHPHHHPHPH $n = 16$



15 contacts

	H	P
H	-1	0
P	0	0

	H	P	N	X
H	-4	0	0	0
P	0	1	-1	0
N	0	-1	1	0
X	0	0	0	0

Biomolecules may have kinetic traps which prevent them from reaching equilibrium within the lifetime of the molecule. Long molecules are often trapped in such meta-stable states during transcription.

Possible solutions are

- Stochastic folding simulations (predict folding pathways)
- Predicting structures for growing fragments of the sequence
- Analysis of the energy landscape based on complete suboptimal folding

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.

RNA

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

dynamic programming algorithms available

Lattice proteins

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

problem is NP-hard

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

- A set X of configurations
- a notion \mathfrak{N} of neighborhood, nearness, distance or accessibility on X , and
- an energy function $f : X \rightarrow \mathbf{R}$

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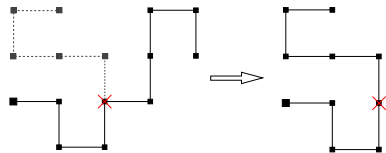
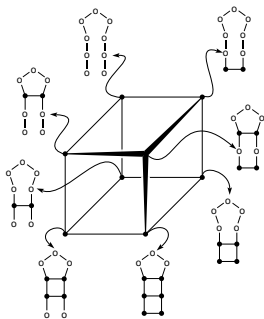
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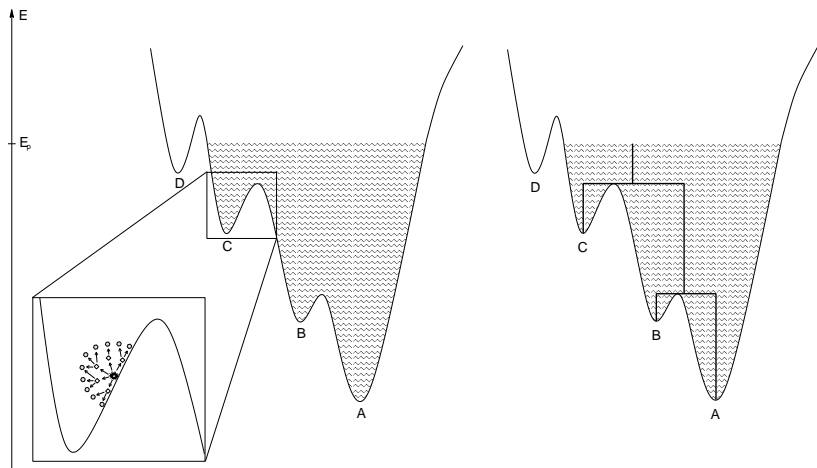
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The move set



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

Low-energy states of lattice proteins



Kinetic Folding Algorithm

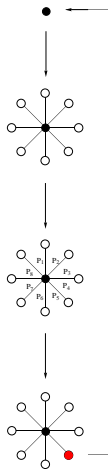
Simulate folding kinetics by a rejection-less Monte-Carlo type algorithm:

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 $1/\sum_i P_i$.

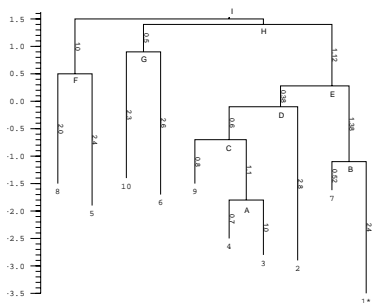


Energy barriers and barrier trees

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 that can be reached by a downhill walk starting at this point

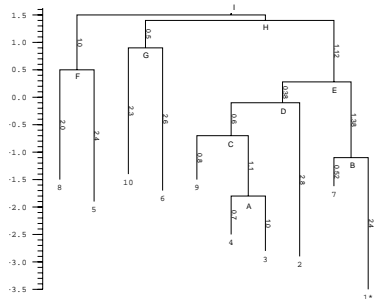


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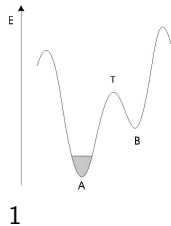
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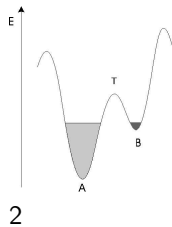
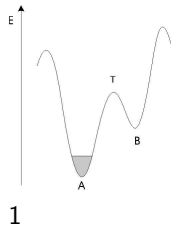
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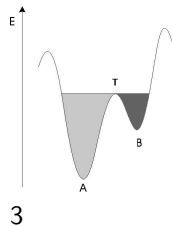
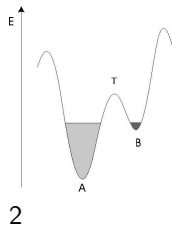
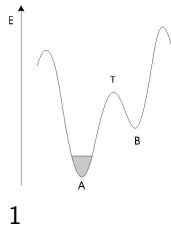
The flooding algorithm



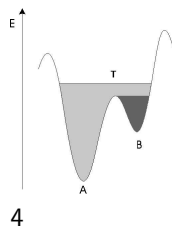
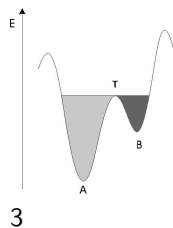
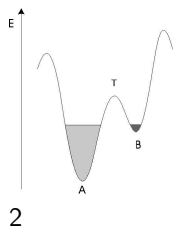
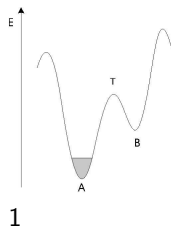
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The flooding algorithm



The flooding algorithm



Information from the barrier trees

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

N.B.: A *gradient basin* is the set of all initial points from which a gradient walk (steepest descent) ends in the same local minimum.

Dynamics of biopolymers

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?)

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

Barrier tree kinetics

For a reduced description we need

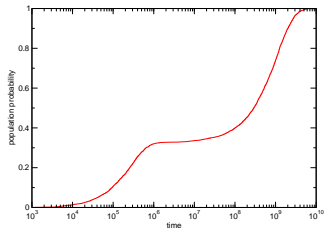
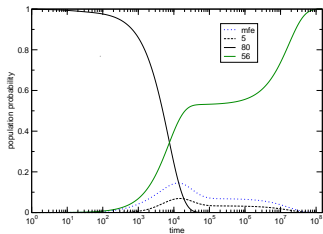
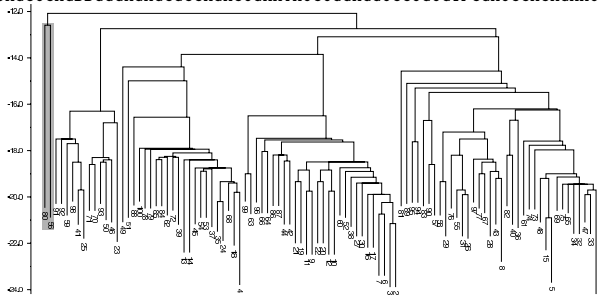
- **macro-states** that form a partition of full configuration space
- **transition rates** between macro-states, e.g.

$$r_{\beta\alpha} = \Gamma_{\beta\alpha} \exp\left(-\frac{(E_{\beta\alpha}^* - G_{\alpha})}{kT}\right)$$

All relevant quantities can be computed via the flooding algorithm.

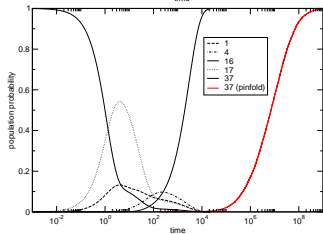
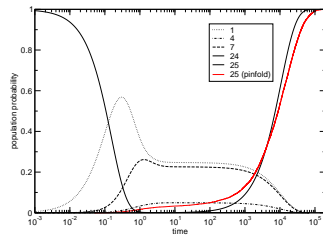
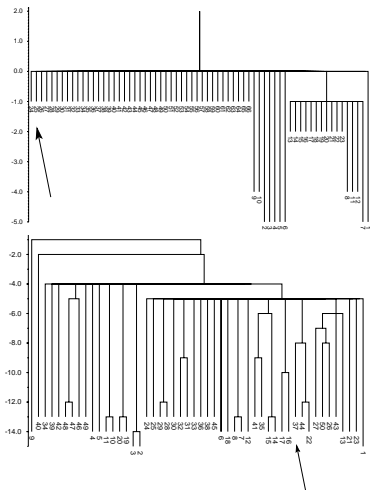
Dynamics of tRNA

CGCGAUUUAGCUCAGDDGGGAGAGCGCCAGACUGAAYAUCUGGAGGUCCUGUGTPCGAUCCACAGAAUUCGCACCA



Dynamics of lattice proteins: HEX/TET lattice

NNHPPNPNPHHHHPXP $n = 16$



Conclusion

- **Discrete models** allow a detailed study of the energy surface.
- **Barrier trees** approximate the landscape topology and folding kinetics.
- A **macrostate approach** of folding kinetics reduces simulation time drastically.
- The **accuracy of the model** is very high in the case of RNA and mostly sufficient for lattice proteins.
- This **newly generated framework** provides a powerful method for further refinement of biopolymer folding landscapes.

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Thanks

Peter Stadler

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the audience