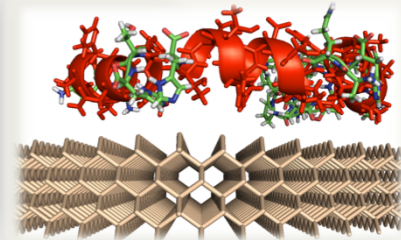
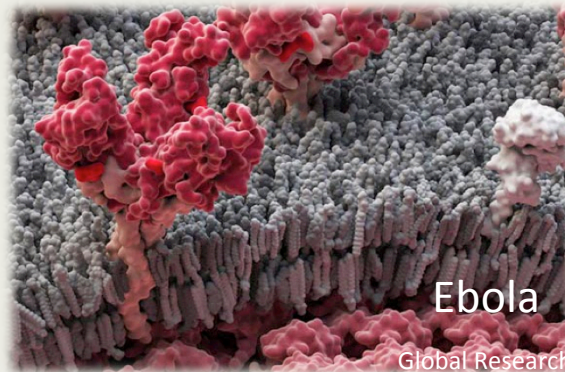

A 20th Century Physics No-no Problem:

The Reasons of Us Being

Michael Bachmann

Soft Matter Systems Research Group, The University of Georgia, Athens GA (USA)

Universidade Federal de Mato Grosso, Cuiabá • Universidade Federal de Minas Gerais, Belo Horizonte (Brazil)

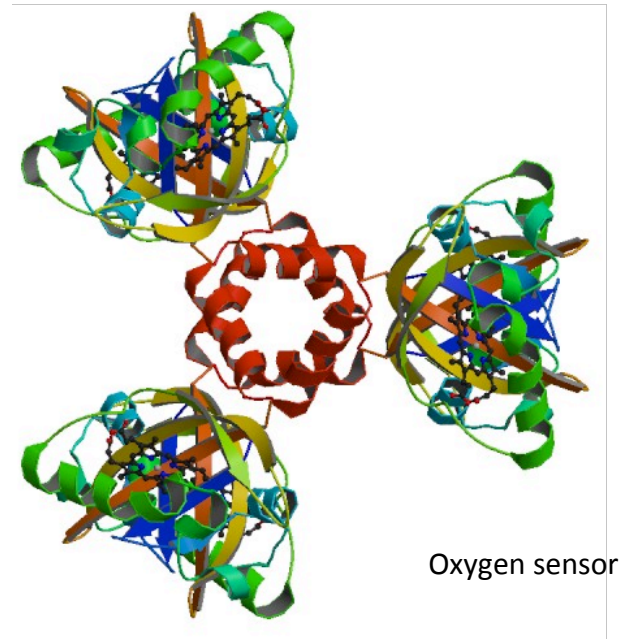


Joint NTZ-SFB/TRR 102 Colloquium, University of Leipzig, 11/27/2014



Overview

1. Introduction: Bioproteins
2. Modeling and Simulation
3. Applications
4. Summary



1. Introduction: Bioproteins

Sequence
↓
Conformation
↓
Function

Ribosomal synthesis of amino acid sequences -
but not of functional geometric structures!

Proteins: Amino acid sequences
encoded in the DNA

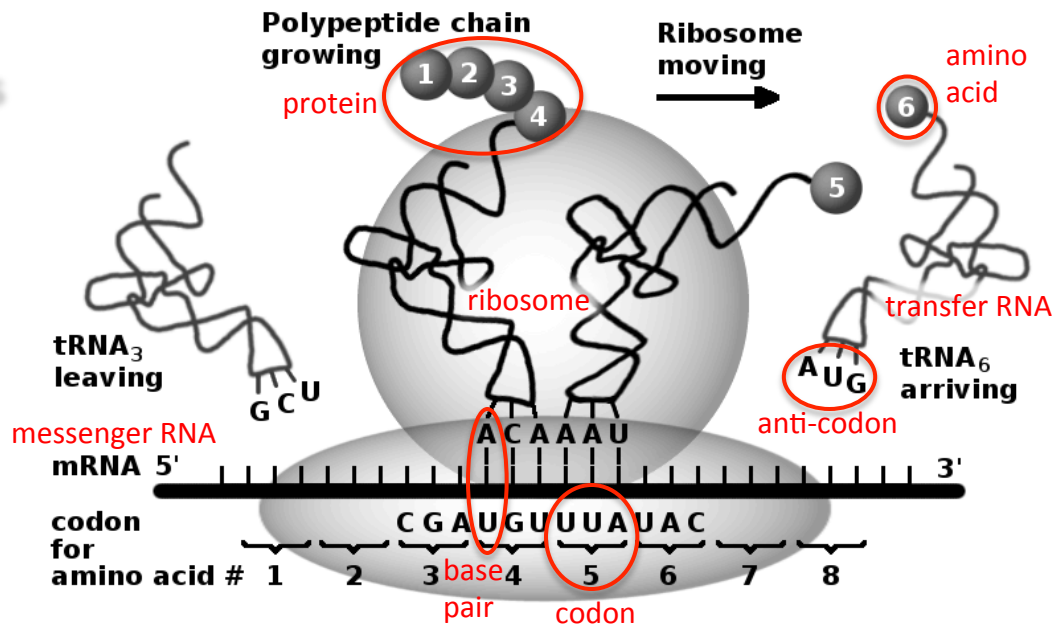
- Polypeptides:**

$N \sim 20 - 27000$
amino acids linked by
peptide bonds

- 20 amino acids:**

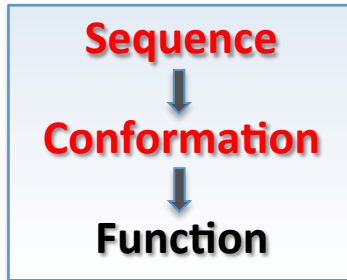
20^N possible sequences

- But:** only $\sim 10^6$ functional
proteins (human)



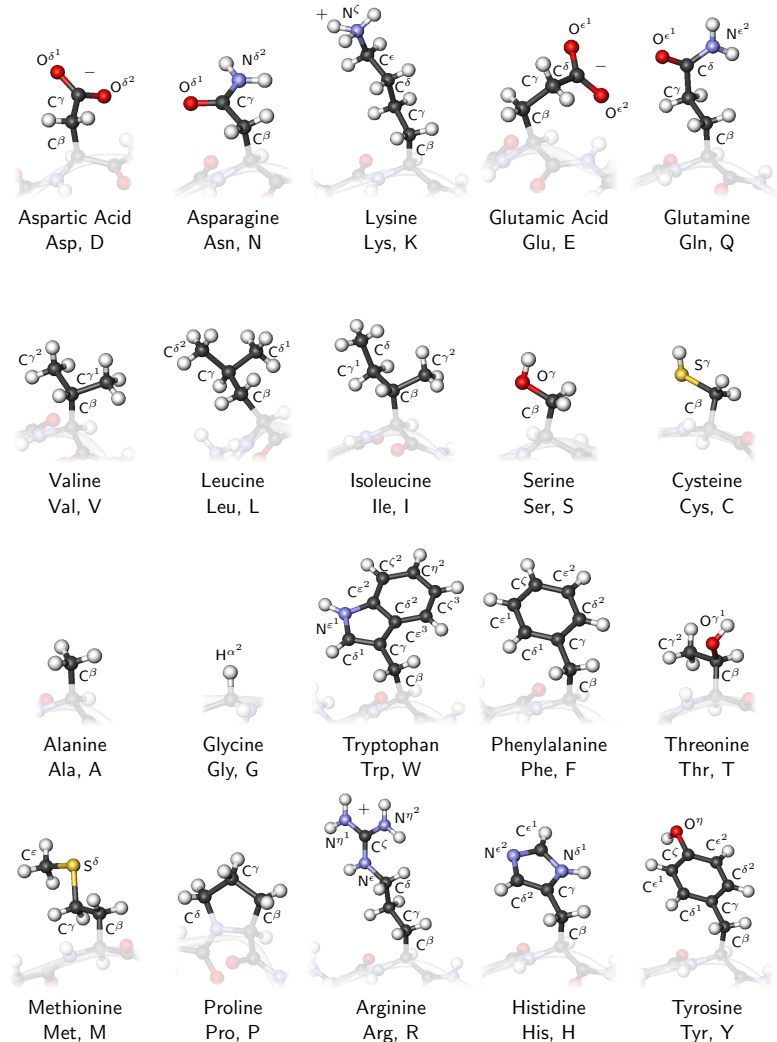
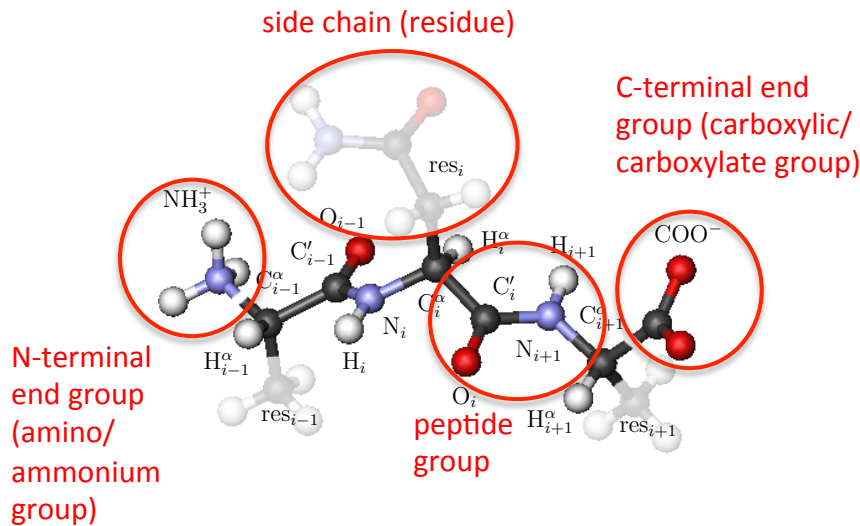
4 base pair combinations, 3 per codon: 64 codes

1. Introduction: Bioproteins

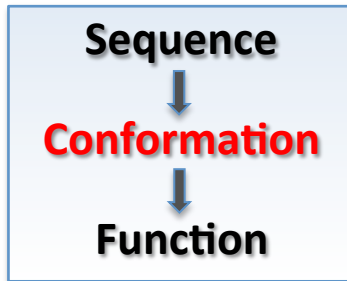


20 typical amino acids side chains

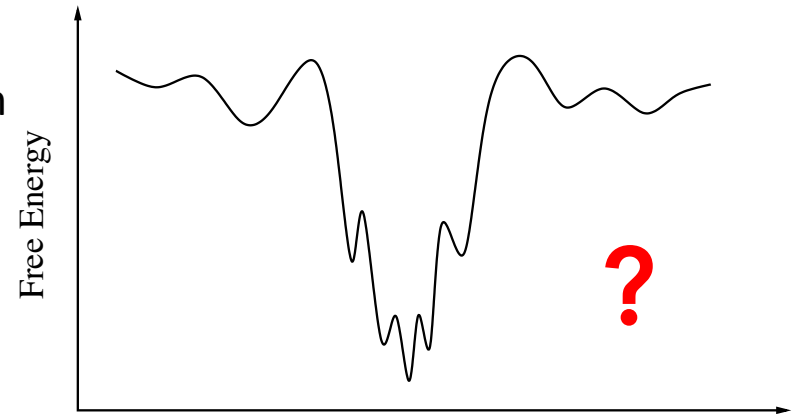
Protein backbone



1. Introduction: Bioproteins

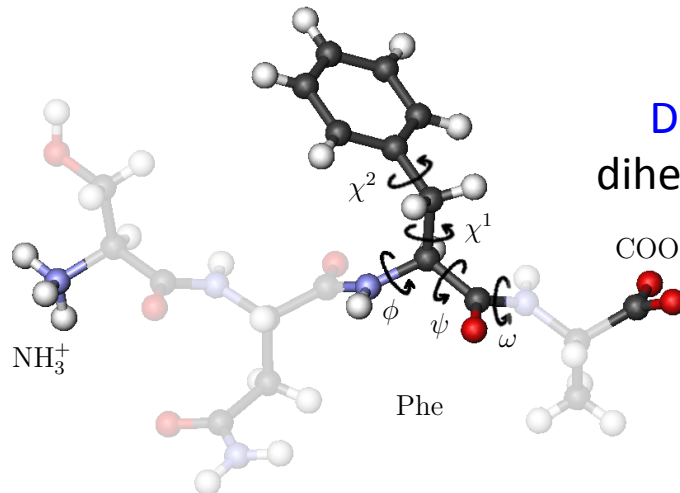


Free-energy landscape:
rugged, funnel-like, with
pockets



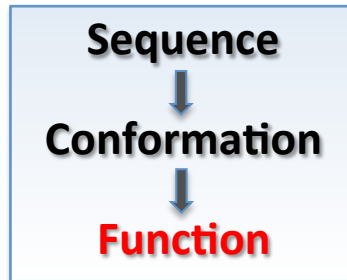
Sequence determines structure.

- Spontaneous folding into native conformation (Anfinsen's experiments)
- Folding pathway(s): free-energy driven stochastic search
- Protein folding time scale: milliseconds to seconds
- Metastability



Degrees of freedom:
dihedral and side-chain
torsion angles

1. Introduction: Bioproteins



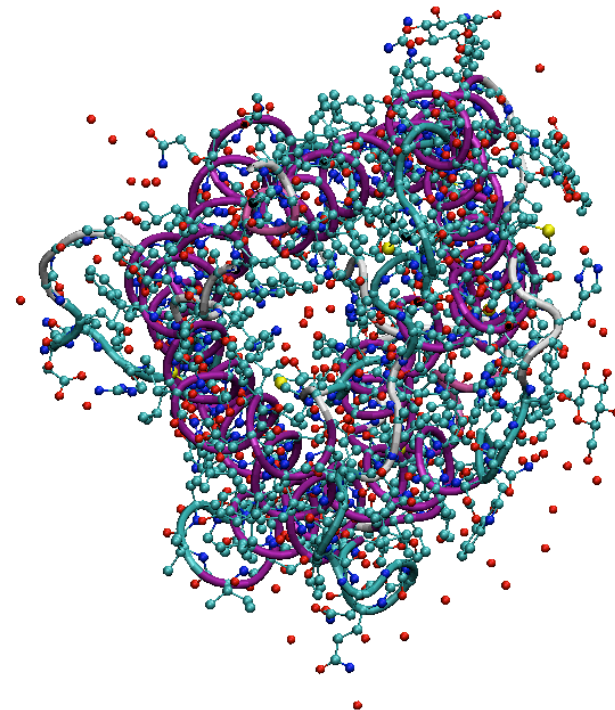
3D protein structure determines functionality.

Example: Aquaporin

transfer of 3,000,000,000 water molecules / sec.

Proteins are involved in virtually all cell processes.

- Transport (channels, pores)
- Cell stability (actin filaments)
- Catalytic activity (enzymes)
- Molecular motors (muscle contraction, DNA polymerase, ATP synthase)
- Epidemic diseases (Alzheimer's, Huntington's, virus infections, etc.)



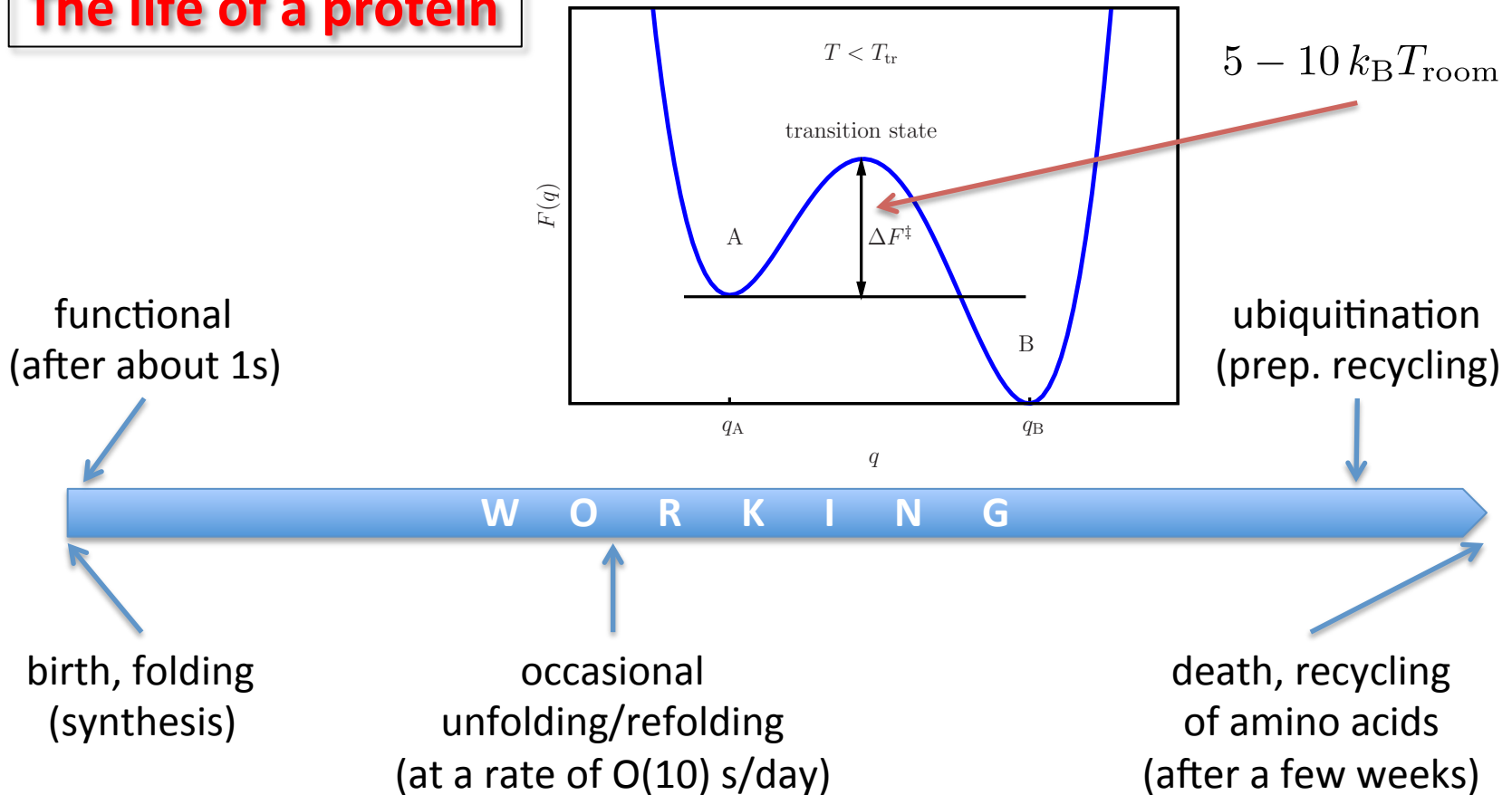
1. Introduction: Bioproteins

Reasons for nature's preference of macromolecular matter:

- ✓ Small energetic margins allow for complex processes on nanometer scales
- ✓ Small number of degrees of freedom (torsion angles), but large number of constraints (bond lengths, bond angles)
- ✓ Sequential disorder (amino acid sequence), large number of possibilities, easy to encode (RNA)
- ✓ Formation of stable, but flexible molecular structures on mesoscopic scales
- ✓ Preference of finite-size first-order-like transitions (phase separation rather than phase coexistence → entropic suppression: “surface entropy”)
- ✓ Broken long-range symmetries, but long-range cooperativity (amorphous, glasslike) in molecular structures
- ✓ Production of identical macromolecules in huge numbers, statistics
- ✓ Process efficiency (signal exchange, transport, catalysis, etc.)
- ✓ Simple strategies for adaptation, regulation, and genetic evolution

1. Introduction: Bioproteins

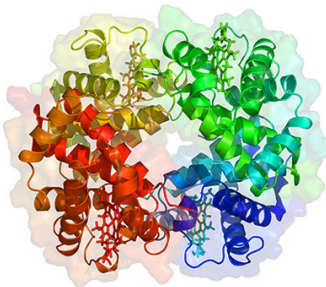
The life of a protein



1. Introduction: Bioproteins

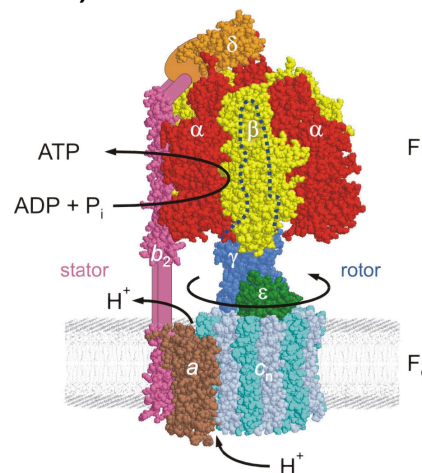
Challenges:

- Single protein: Heterogeneous chain composed of up to 27,000 amino acids
- Geometric structure \leftrightarrow Biological function
- Structure formation process \leftrightarrow Structural phase transition on finite scales?
- But: no thermodynamic limit, no scaling, no (unique) transition points
- Finite-size, surface, and disorder effects



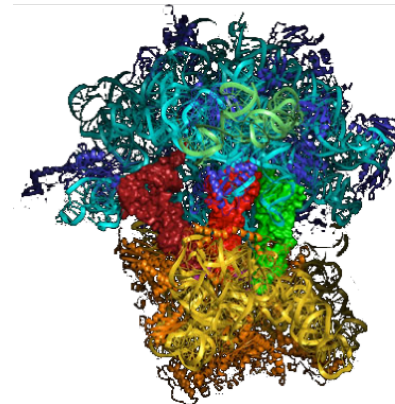
Hemoglobin

Oxygen transport in red blood cells, 4 units, >2,000 atoms



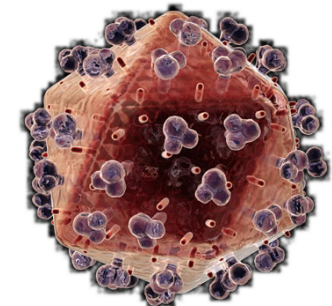
ATP synthase

Production rate 100 ATP/s, 2 units, 40,000 atoms



Ribosome

Protein synthesis, 2 units, 200,000 atoms



Herpes

Icosahedral virus, 10^5 - 10^6 atoms

1. Introduction: Bioproteins

The Dark Side: Proteomic Diseases

- **Acquired Immuno-Deficiency Syndrome** (**H**uman **I**mmunodeficiency **V**irus)
>30 million infected people worldwide (WHO); incurable
- **Hepatitis B** (**H**epatitis **B** **V**irus)
2 billion concerned, >350 million chronic (WHO), incurable
- **Bovine Spongiform Encephalopathy** (prionic disease, cattle)
US beef export: 2003: \$3.8 billion, 2005: \$1.4 billion (USA Today)
- **Creutzfeld-Jakob Disease** (prionic, human)
one sick per 1 million people; incurable
- **Alzheimer's Disease** (amyloid peptide cluster)
65 years old: 2-3%, at the age of 85: 25-50%
24 million sick worldwide (AFI e.V.); incurable
- **Influenza** (virus)
bird flu, swine flu, ...
- **Ebola** (virus)
death rate 50-90% (2014: >1,500), incurable

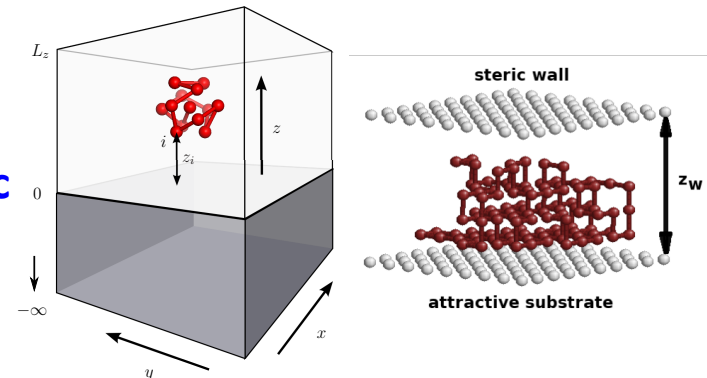
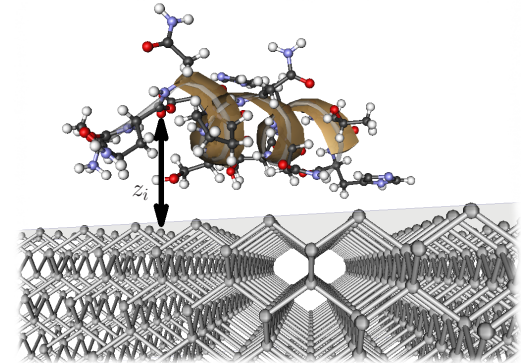


2. Modeling and Simulation

Basic problem: Mesoscopic length scales, 10^1 - 10^3 nm (proteins)

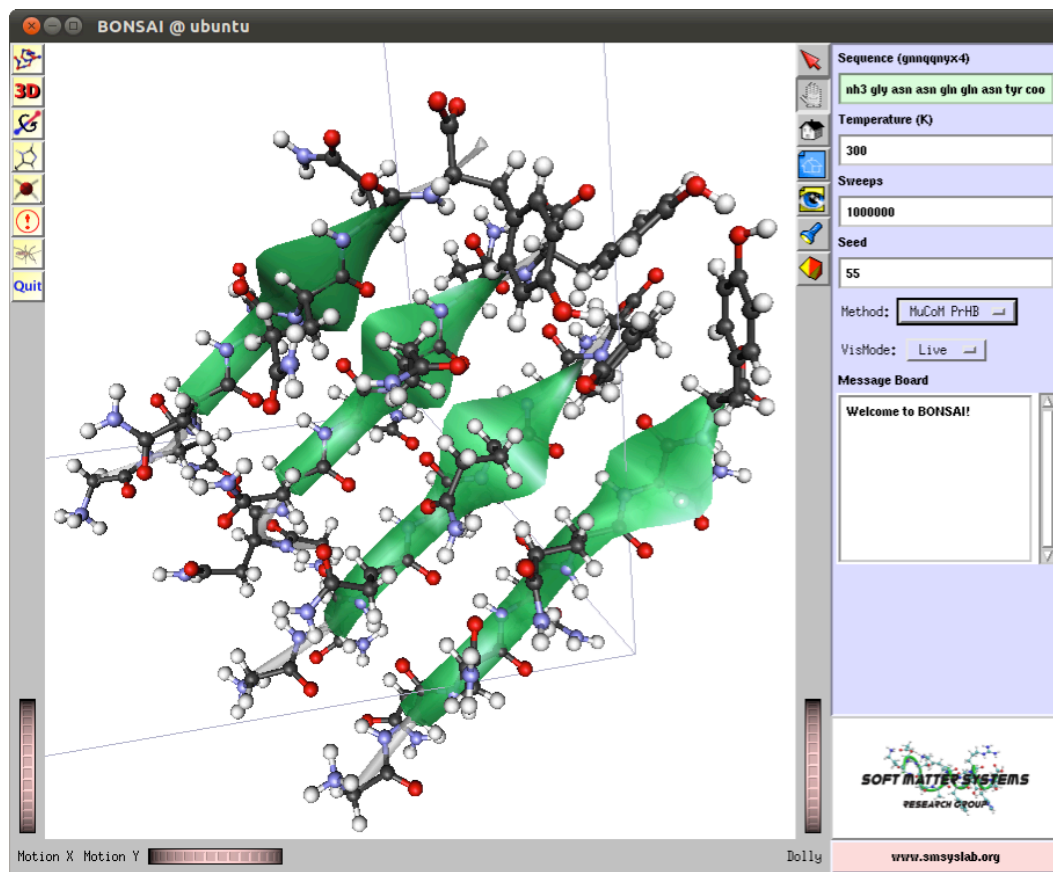
Two major approaches to macromolecular modeling:

- **Semi-empirical, semi-classical all-atom models**
 10^2 - 10^3 parameters (!)
attempting to model short- and long-range interactions (including quantum effects) and solvent effects as accurately as possible, aiming at **specific questions** (e.g., short-time dynamics of folding and function: ns to μ s)
- **Coarse-grained, effective monomeric models**
< 10 parameters
extreme abstraction, concentration on **generic properties** such as structural transitions
→ only way to study **complexity**



2. Modeling and Simulation

All-atom models (using simulation packages: CHARMM, Amber, Gromacs, etc.) ...



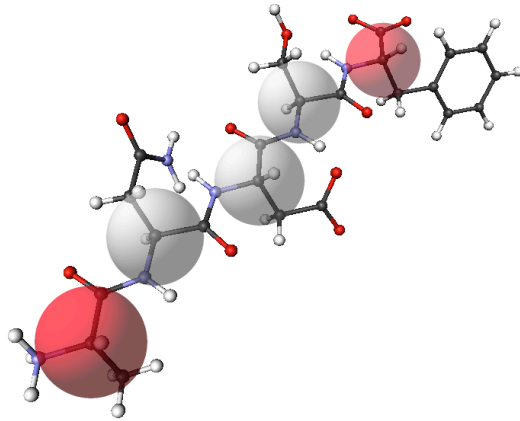
Monte Carlo Sim Pack

- ⇒ Protein folding
- ⇒ Protein aggregation
- ⇒ Protein adsorption

- ⇒ C/C++ modularity
- ⇒ OpenInventor, Coin3D
- ⇒ Multiple sim threads

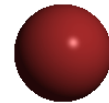
2. Modeling and Simulation

... or coarse-grained models for effective non-bonded interactions:

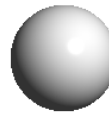


“Hydrophobic-polar” proteins:

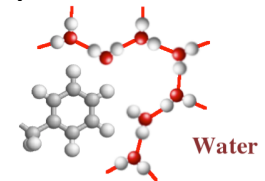
20 amino acids \Rightarrow 2 major groups:



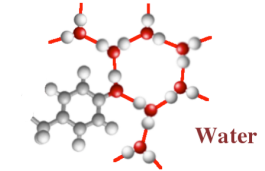
Hydrophobic
(water-repellent)



Polar, hydrophilic

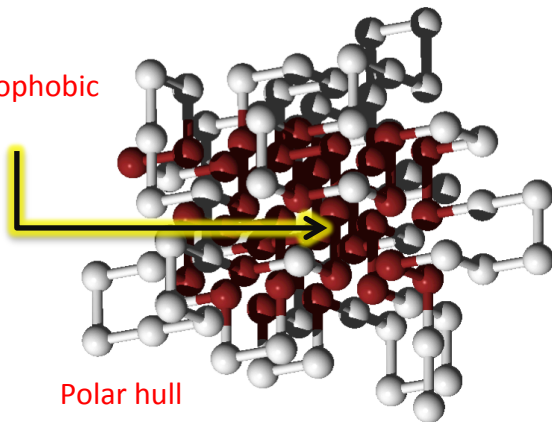


Phe



Tyr

Hydrophobic core



Polar hull

HP model

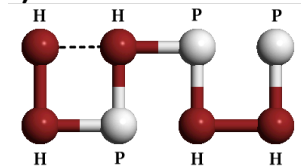
(“Ising model of lattice proteins”):

$$E(\mathbf{X}) = -\varepsilon_{HH} n_{HH}(\mathbf{X})$$

\mathbf{X} conformation (coordinates)

ε_{HH} H-H energy scale

n_{HH} number of H-H contacts



[lattice (HP model): K. F. Lau, K. A. Dill, *Macromolecules* **22**, 3986 (1989);

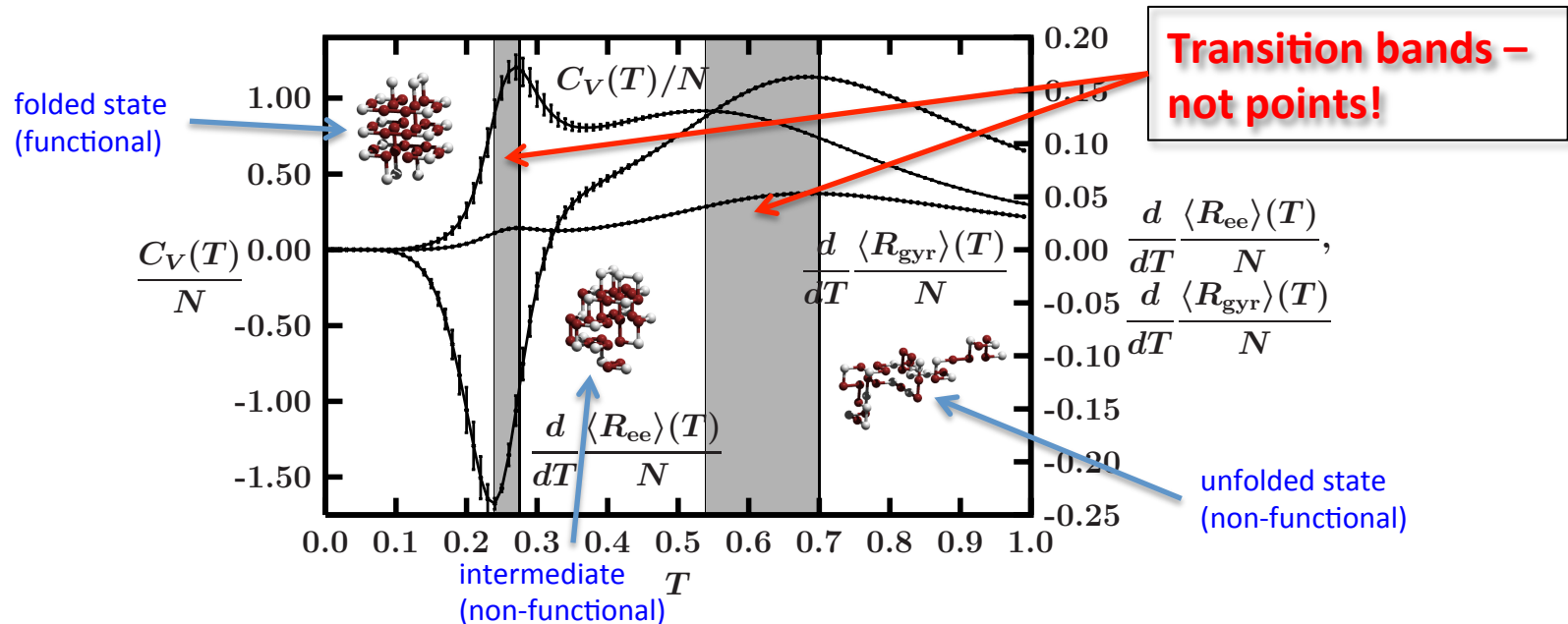
off-lattice (AB model): F. Stillinger, T. Head-Gordon, C. L. Hirshfeld, *Phys. Rev. E* **48**, 1469 (1993).]

2. Modeling and Simulation

HP model (and other toy models for proteins)

- cannot address specific questions about proteins
- but:** some value for studies of more general aspects of macromolecular structure formation

Thermodynamics of protein folding



[M.B., W. Janke, Phys. Rev. Lett. **91**, 208105 (2003); J. Chem. Phys. **120**, 6779 (2004).]

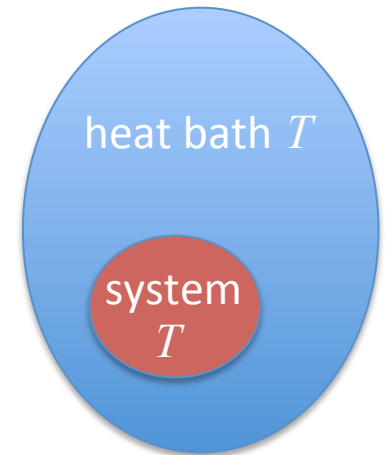
2. Modeling and Simulation

Transition bands: Undesired ambiguity in **temperature** space

What is temperature?

typically introduced in *macroscopic environments* as:

- an easy-to-control **external parameter** in experiment
- a **canonical equilibrium property** of a system in a heat bath
- a **constant in Boyle-Mariotte's ideal gas law** (17th century!!):
isothermal process $pV = \text{const}$ (later Clapeyron: $pV = nRT$)
⇒ measured with thermometers



So, asking more precisely: **What is the temperature of a finite system?**

- can use the **microcanonical equilibrium property** of two systems in contact:

$$T_1 = \left(\frac{\partial S_1(E_1)}{\partial E_1} \right)^{-1} = \left(\frac{\partial S_2(E_2)}{\partial E_2} \right)^{-1} = T_2$$

S_i microcanonical entropy
 E_i energy

Important: Microcanonical temperature is system property!

2. Modeling and Simulation

A somewhat flaky proof



total density of states (multiplicative?): $g(E_1, E_2) = g_1(E_1)g_2(E_2)$

total entropy (additive?): $S(E) = k_B \ln g(E) \Rightarrow S(E_1, E_2) = S_1(E_1) + S_2(E_2)$

total energy (additive?): $E = E_1 + E_2 \Rightarrow dE = dE_1 + dE_2$

equilibrium: $dS = 0, dE = 0$

$$\begin{aligned}\text{Thus: } dS(E_1, E_2) &= \left(\frac{\partial S}{\partial E_1} \right)_{E_2} dE_1 + \left(\frac{\partial S}{\partial E_2} \right)_{E_1} dE_2 \\ &= \frac{\partial S_1}{\partial E_1} dE_1 + \frac{\partial S_2}{\partial E_2} dE_2 = \left(\frac{\partial S_1}{\partial E_1} - \frac{\partial S_2}{\partial E_2} \right) dE_1 \\ &= 0\end{aligned}$$

$$\text{and: } \frac{\partial S_1(E_1)}{\partial E_1} = \frac{\partial S_2(E_2)}{\partial E_2} \equiv \frac{\partial S(E)}{\partial E} \Rightarrow T_1(E_1) = T_2(E_2) \equiv T(E)$$

(only valid for dilute gas, neglecting interactions); **we define:** $T(E) := \left(\frac{\partial S(E)}{\partial E} \right)^{-1}$

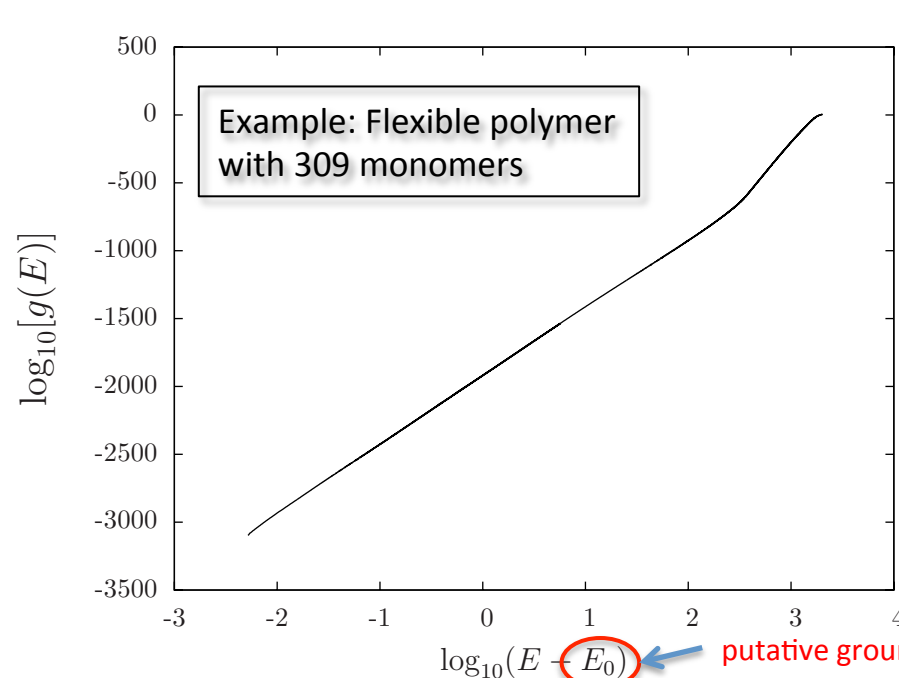
2. Modeling and Simulation

Microcanonical entropy and density of states

Systematic identification and classification of transitions in finite systems:

“Back to the basics”: **Energy and entropy govern cooperative behavior of a system.**

Microcanonical entropy: $S(E) = k_B \ln g(E)$, where $g(E)$ is the **density of states**:



$$g(E') = \int \mathcal{D}\mathbf{X} \delta(E' - E(\mathbf{X}))$$

No mathematical calculation possible for almost all complex problems

Computational challenge:

Accurate estimation of $g(E)$ over **hundreds to thousands of orders of magnitude!!**

2. Modeling and Simulation

Microcanonical inflection-point analysis

inverse temperature

$$\beta(E) = T^{-1}(E) = \frac{dS(E)}{dE}$$

$$\gamma(E) = \frac{d^2 S(E)}{dE^2}$$

Transitions:

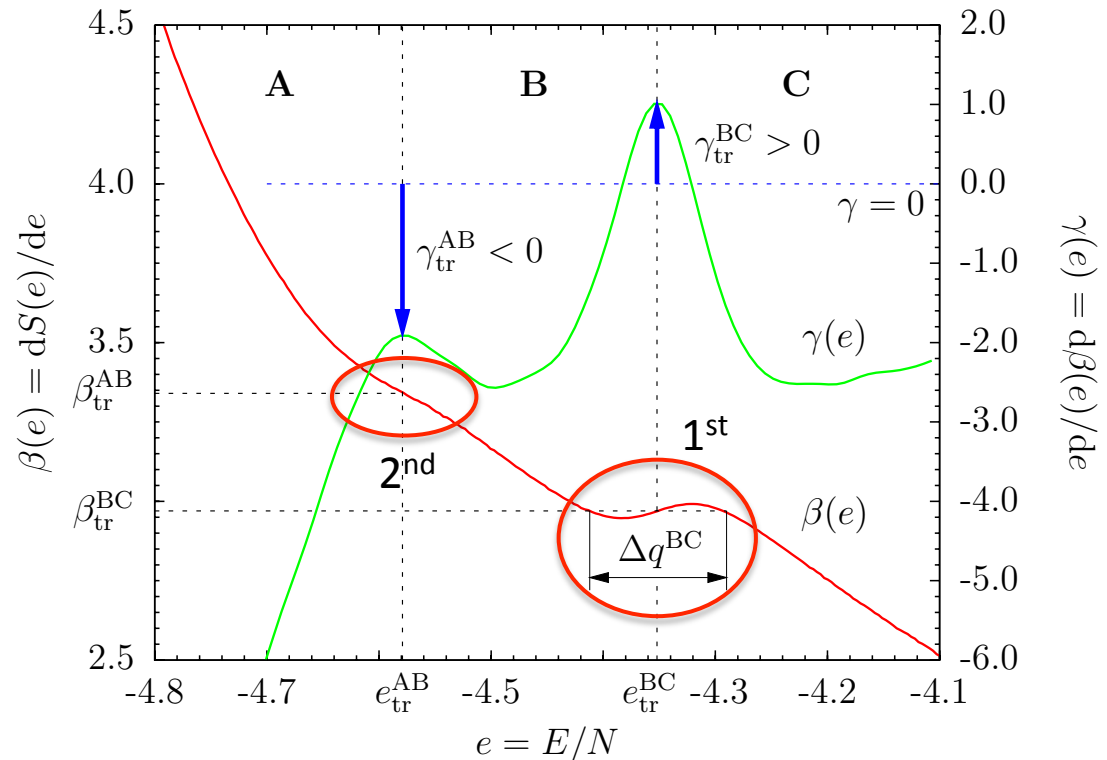
1st order:

$$\gamma_{\text{tr}} > 0 \quad (\Delta q_{\text{tr}} > 0)$$

2nd order:

$$\gamma_{\text{tr}} < 0 \quad (\Delta q_{\text{tr}} = 0)$$

Δq_{tr} latent heat



Generalization possible: Principle of minimal sensitivity $S^{(n)}(E) = \frac{d^n S(E)}{dE^n} = 0$

2. Modeling and Simulation

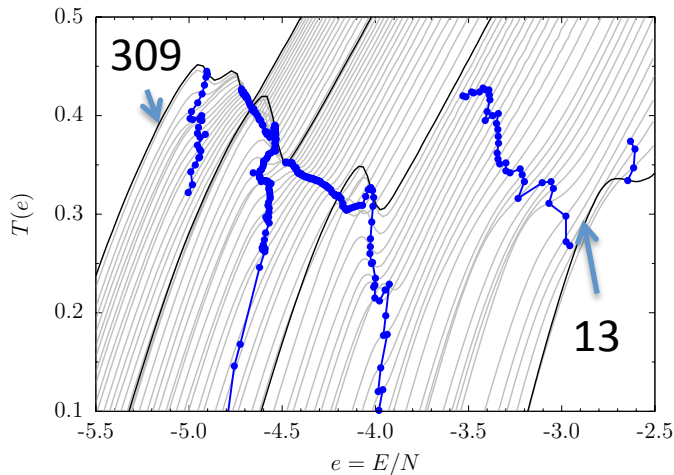
Advanced Monte Carlo simulation techniques:

- Replica exchange (Swendsen, Wang, 1986)
- Multiple-histogram reweighting (Ferrenberg, Swendsen, 1989)
- Parallel tempering (Geyer, 1991; Hukushima, Nemoto, 1996)
- Simulated tempering (Marinari, Parisi, 1992; Lyubartsev et al., 1992)
- Multicanonical sampling (Berg, Neuhaus, 1992)
- Wang-Landau Method (Wang, Landau, 2001)
- ...

Output: Density of states

3. Applications: Nucleation

Crystallization of flexible, elastic homopolymers

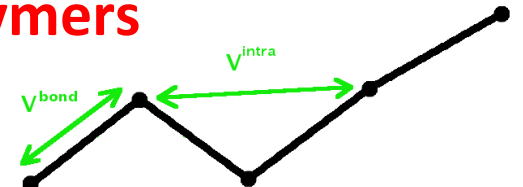
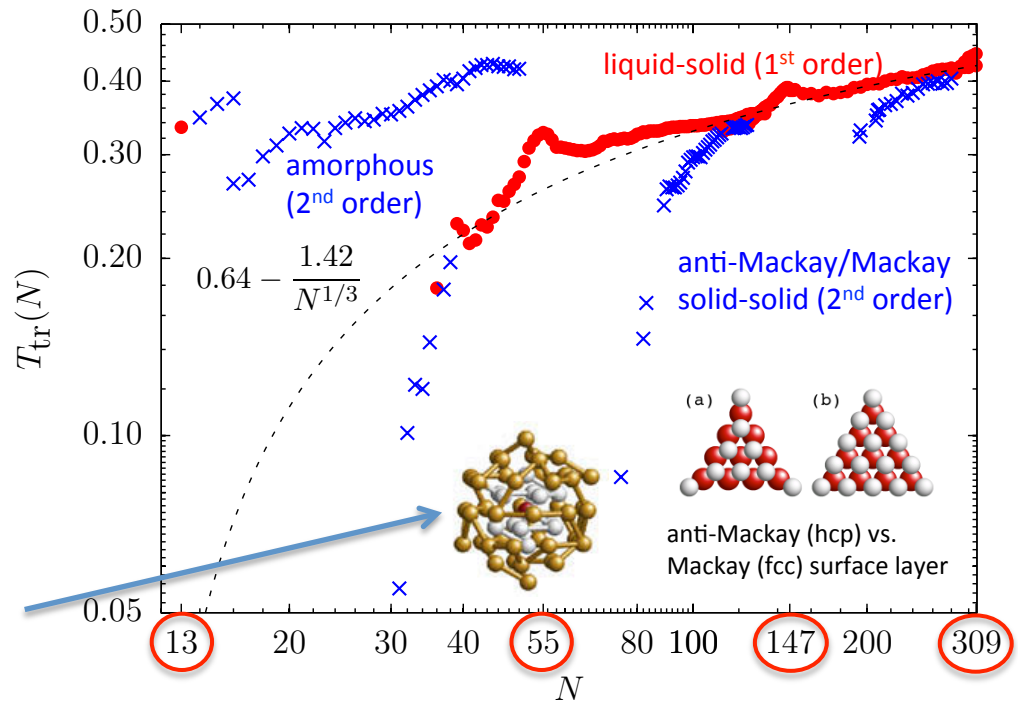


inflection points in low-energy regions

$N = 13 \dots 309$ monomers

icosahedral geometry preferred; perfect for $N=13, 55, 147, 309, \dots$ ("magic" numbers)

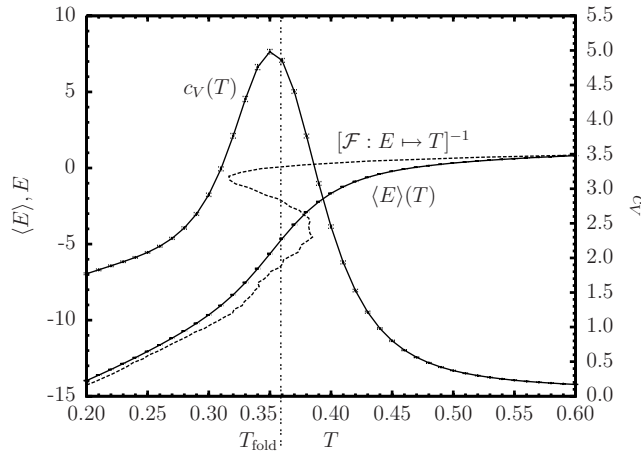
microcanonical transition temperatures



3. Applications: Protein Folding

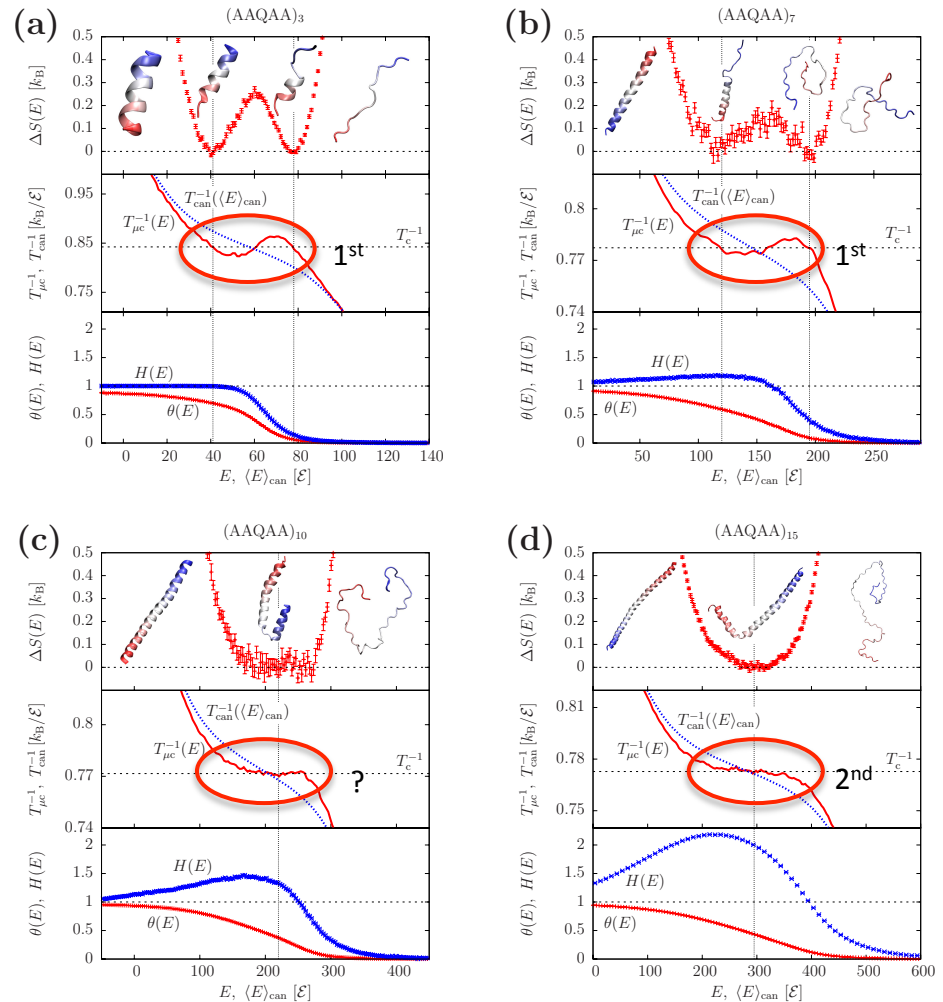
Peptide folding

typical 1st order scenario:



Helix-coil transition of peptide $(AAQAA)_n$ with $n=3, 7, 10, 15$:

Transition changes from 1st order to 2nd order for longer chains (not preferred in natural proteins)



3. Applications: Protein Folding

Helix-bundle formation

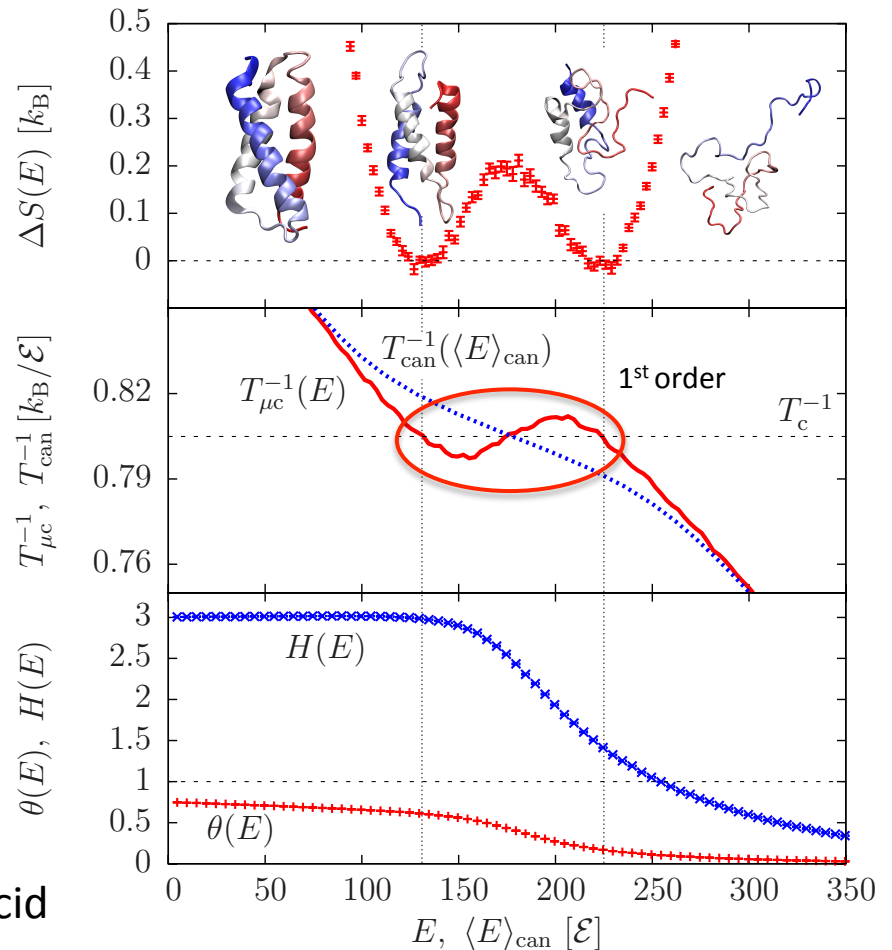
α 3D: three-helix bundle
73 amino acids

clear **first-order signal** of cooperative helix-coil/ tertiary bundle transition

nucleation due to **loss of translational entropy** of random/helical segments: entropic suppression

$H(E)$ average number of helices

$\theta(E)$ helicity of individual amino acid

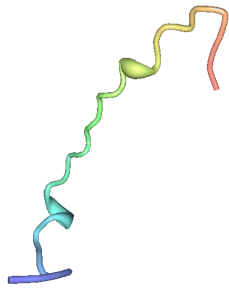


[T. Bereau, M.B., M. Deserno, JACS **132**, 13129 (2010); T. Bereau, M. Deserno, M.B., Biophys. J. **100**, 2764 (2011).]

3. Applications: Protein Aggregation

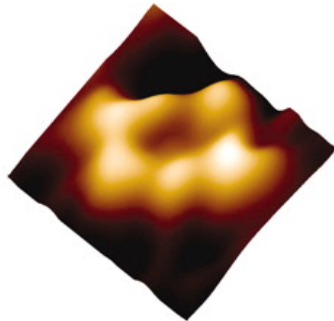
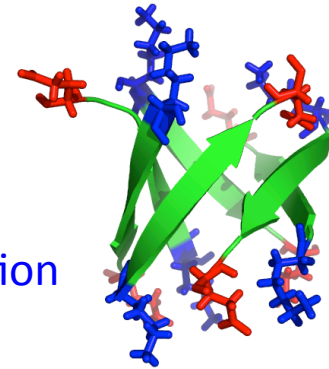
Aggregation of peptides

Alzheimer's disease: Toxicity of misfolded proteins (amyloid hypothesis)



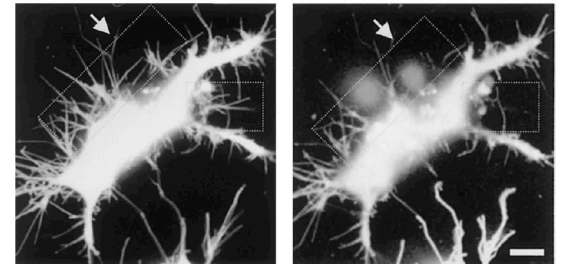
Folding of peptide $A\beta_{1-42}$

Local unfolding, aggregation



Fusion into neural cell membrane, pore formation

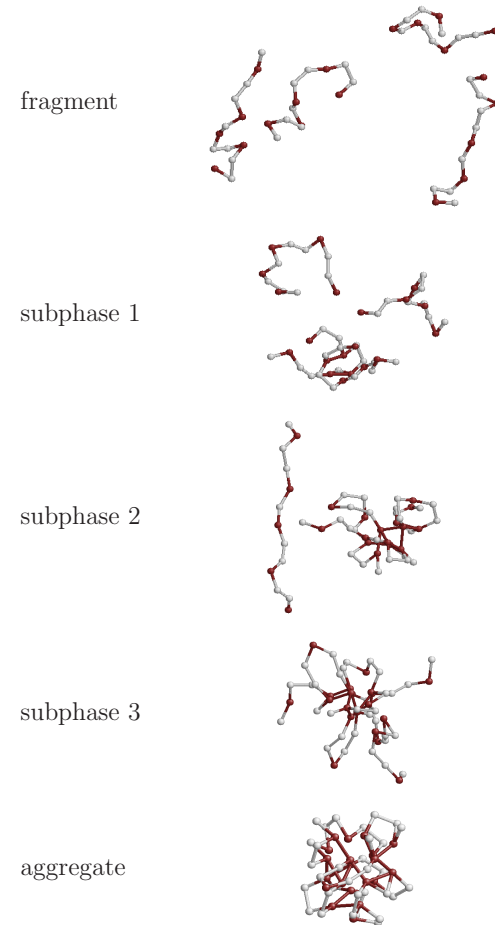
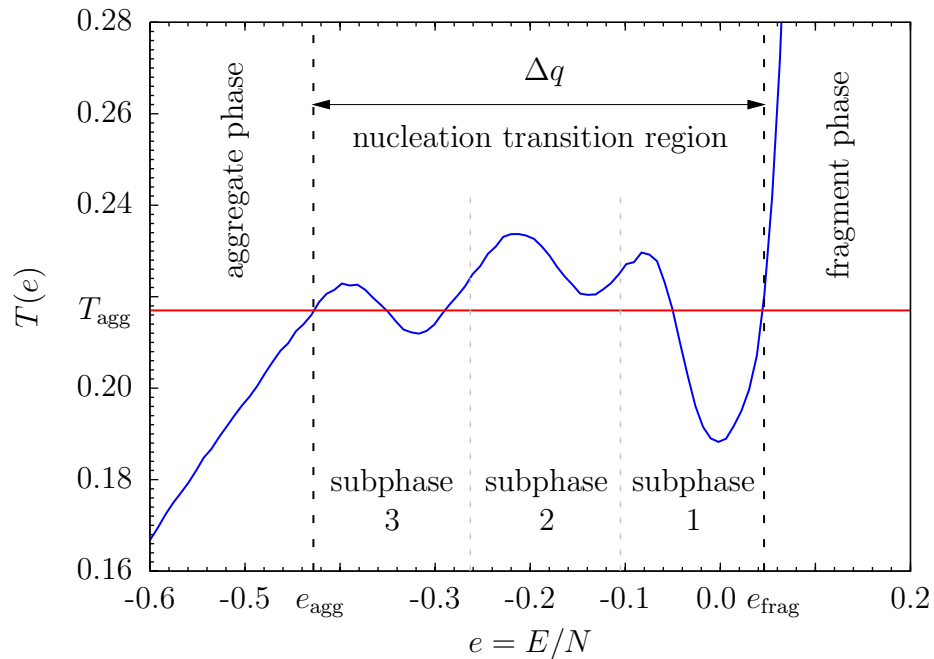
Degeneration of neurons by Ca^{2+} ions



[H. Lin et al., FASEB J. **15**, 2433 (2001); A. Quist et al., PNAS **102**, 10427 (2005); H. A. Lashuel, P. T. Lansbury jr., Quart. Rev. Biophys. **39**, 167 (2006); S. Tomaselli et al., ChemBioChem **7**, 257 (2006); S. Mitternacht, A. Irbäck, Proteins **71**, 207 (2008).]

3. Applications: Protein Aggregation

Example: Aggregation of 4 peptide chains

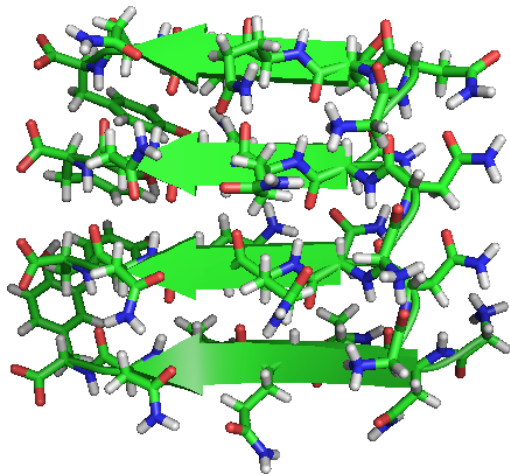


- hierarchy of subphase transitions
- latent heat does not vanish (1st order transition)

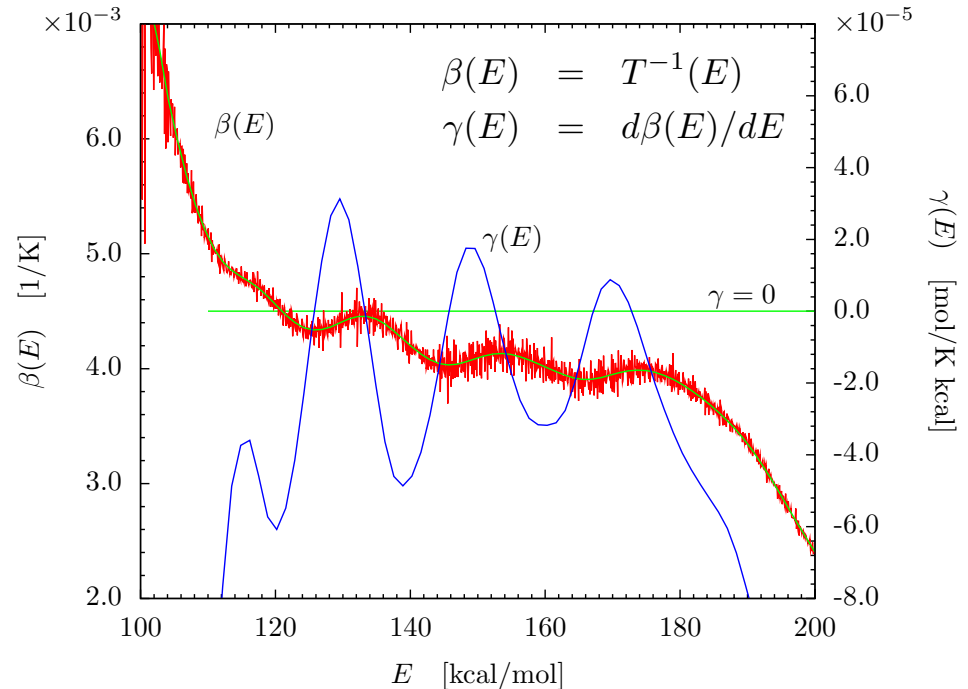
[C. Junghans, M.B., W. Janke, Phys. Rev. Lett. **97**, 218103 (2006); J. Chem. Phys. **128**, 218103 (2008); C. Junghans, W. Janke, M.B., Comp. Phys. Commun. **182**, 1937 (2011).]

3. Applications: Protein Aggregation

Protein-protein aggregation: GNNQQNY (4 chains)



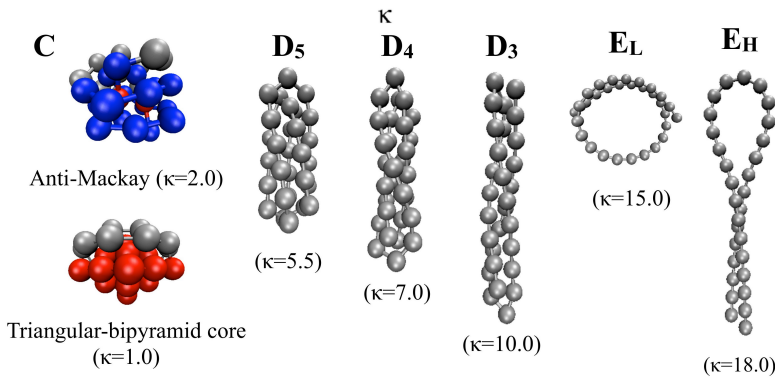
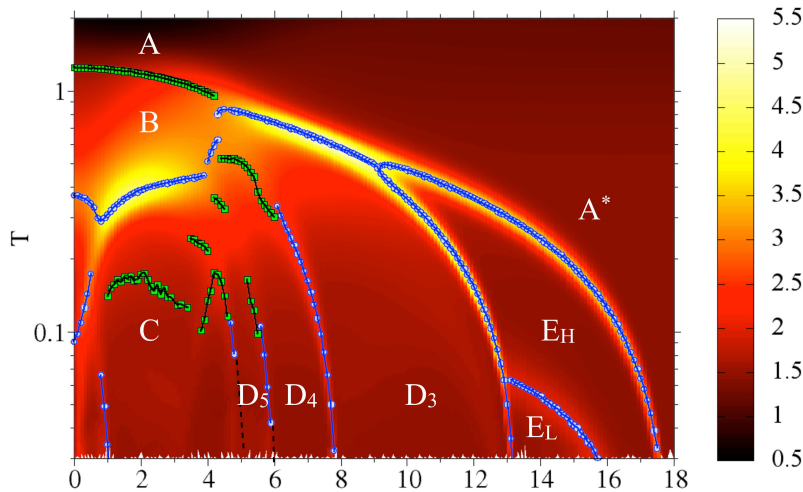
lowest-energy state: β - sheet



- same behavior: **hierarchy of subphase transitions**
- **$O(10)$ CPU years** using parallel tempering Monte Carlo

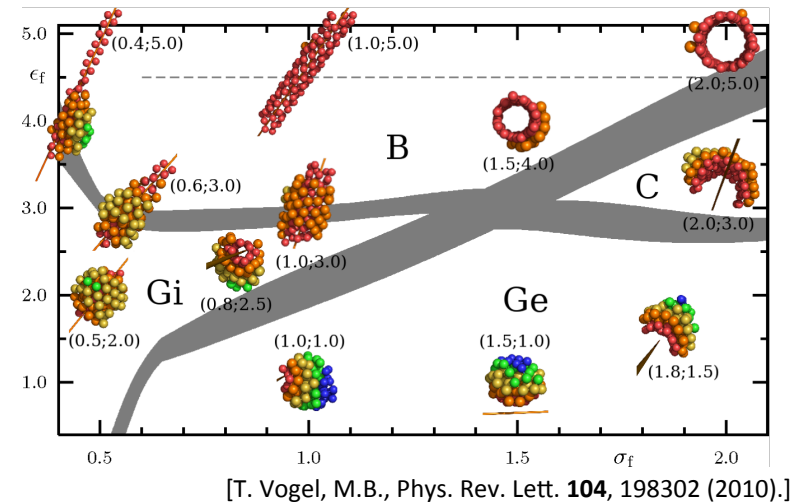
3. Applications: Structural Phase Diagrams

Semiflexible polymers



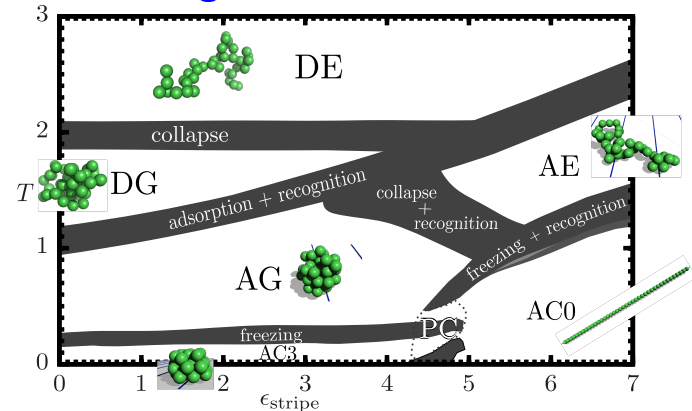
[D. T. Seaton, S. Schnabel, D. P. Landau, M.B., Phys. Rev. Lett. **110**, 028103 (2013).]

Adsorption at nanowires



[T. Vogel, M.B., Phys. Rev. Lett. **104**, 198302 (2010).]

Pattern recognition



[M. Möddel, W. Janke, M.B., Phys. Rev. Lett. **112**, 148303 (2014).]

4. In the End...

Summary:

- Structural transitions of macromolecules exhibit features resembling **thermodynamic phase transitions**.
- Transitions are dominated or at least influenced by **finite-size and surface effects**.
- Studies require **sophisticated algorithmic approaches** and advanced **computational resources** (which is why it is a 21st century problem!).
- Novel or rediscovered **statistical physics methods** enable efficient and systematic analyses.
- Studies of **transitions on mesoscopic scales** help understand phase transitions in macroscopic systems even better.

Ads:



- Website: www.smsyslab.org
- M.B., *Thermodynamics and Statistical Mechanics of Macromolecular Systems* (Cambridge Univ. Press, 2014)

