

PROTEIN FOLDING: A PARADIGM FOR SOLVING HARD PROBLEMS IN BIOLOGY

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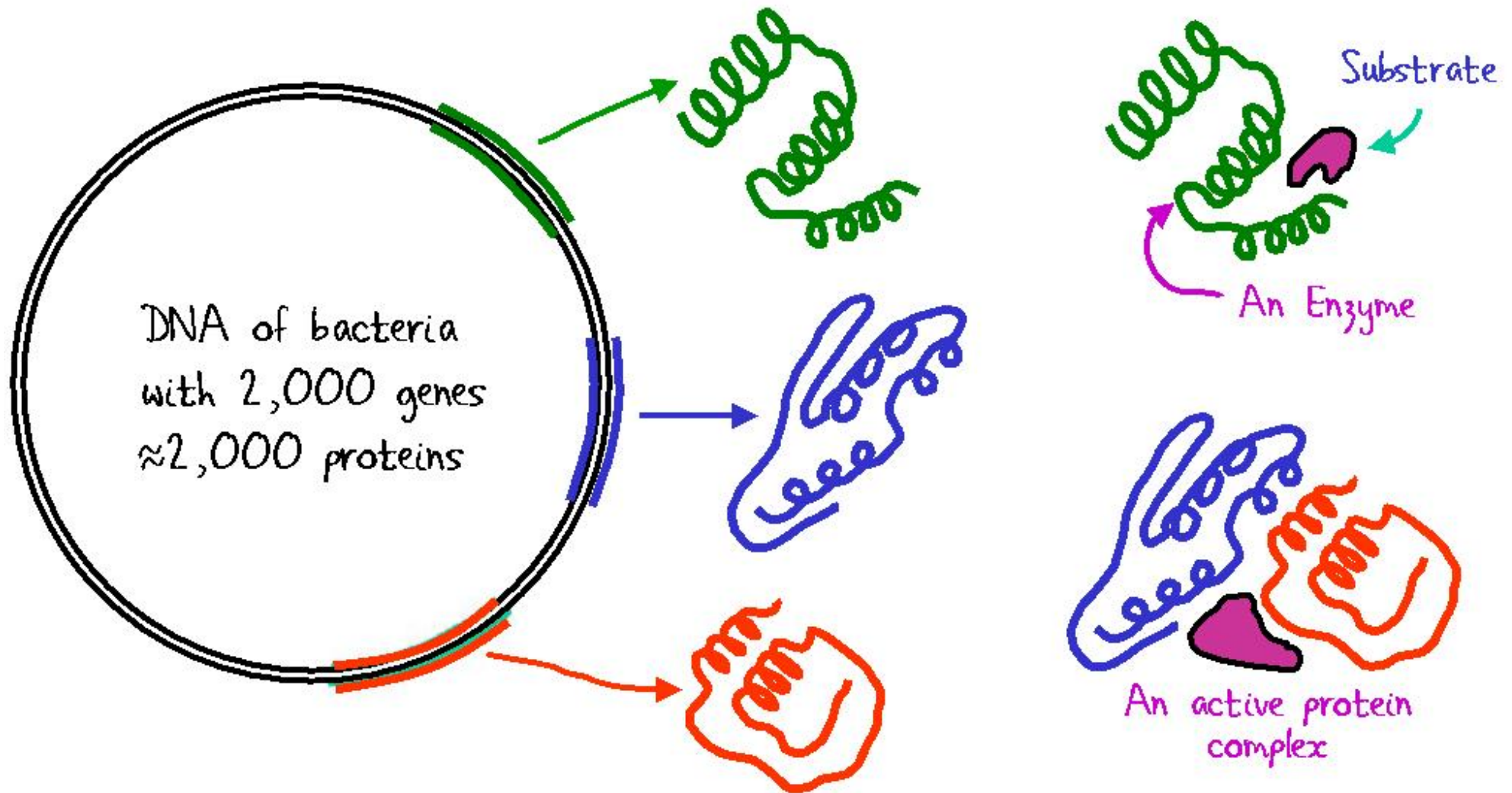
<http://csb.stanford.edu/levitt>

OUTLINE

- Simulation
 - Basic methods
 - Hydrophobic effect
 - Unfolding, folding
- Prediction
 - Special potentials
 - Minimization
 - Monte Carlo
- Hard Problems

INTRODUCTION

STRUCTURAL OVERVIEW OF BIOLOGY



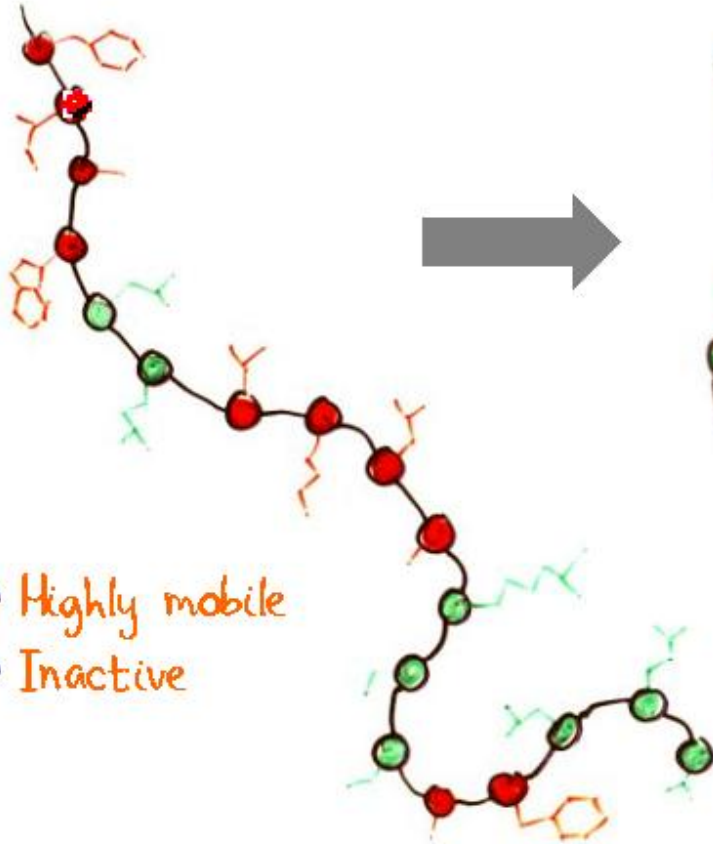
- From DNA sequence, predict all protein structures

- From protein structures predict all function.

PROTEIN FOLDING IS CENTRAL

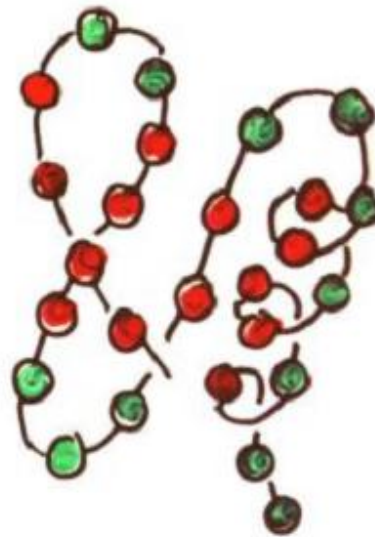
Sequence \rightarrow Structure \rightarrow Function

- Unfolded protein is a chain of amino acids



- Highly mobile
- Inactive

- Folded protein



- Almost unique shape
- Precisely ordered
- Stable
- Active

- Function depends on protein shape



- Specific associations
- Precise reactions

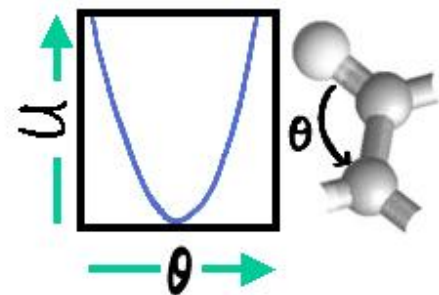
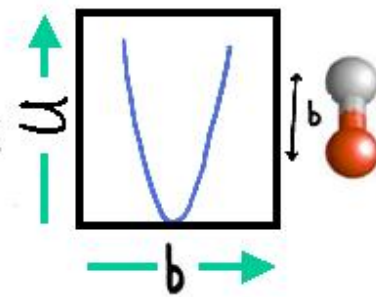
SIMULATION

TOTAL POTENTIAL ENERGY

$$U = \sum \frac{1}{2} K_b (b - b_0)^2 + \sum \frac{1}{2} K_\theta (\theta - \theta_0)^2$$

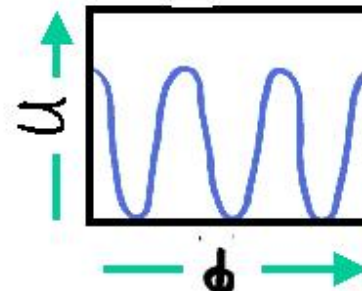
All Bonds

All Angles



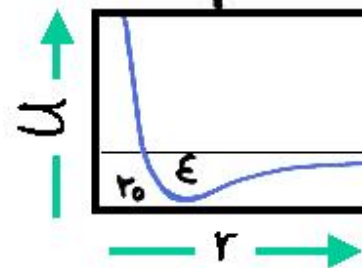
$$+ \sum K_\phi [1 - \cos(n\phi + \delta)]$$

All Torsion Angles



$$+ \sum \epsilon \left[\left(\frac{r_0}{r} \right)^{12} - 2 \left(\frac{r_0}{r} \right)^6 \right]$$

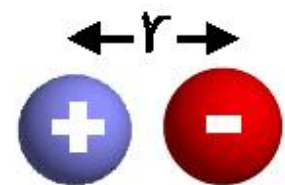
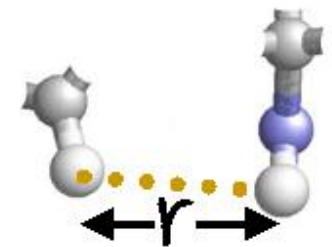
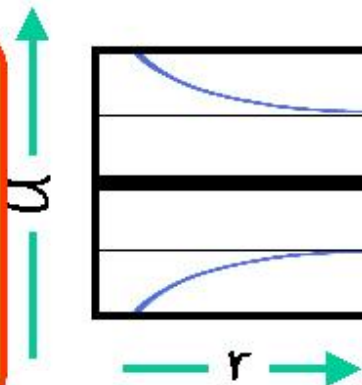
All nonbonded pairs



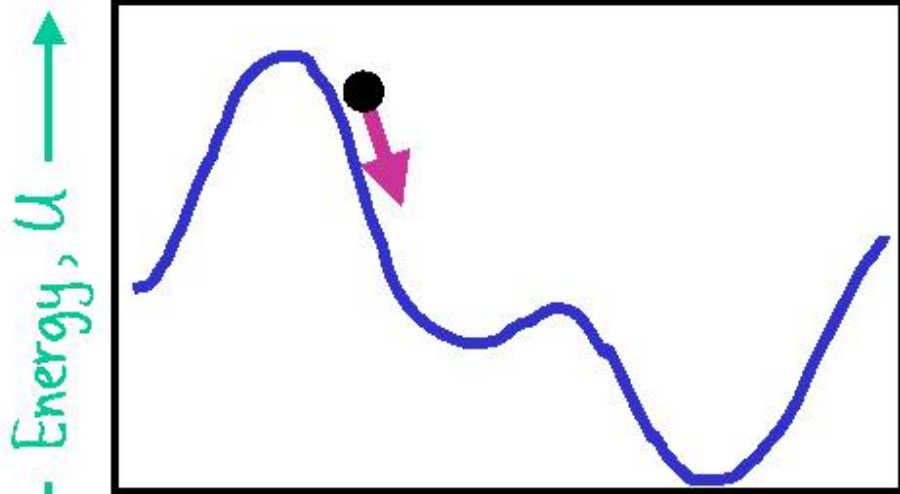
$$+ \sum \frac{332 q_i q_j}{r}$$

All partial charges

ENCAD.
Parameters
from 1979
(Lifson)

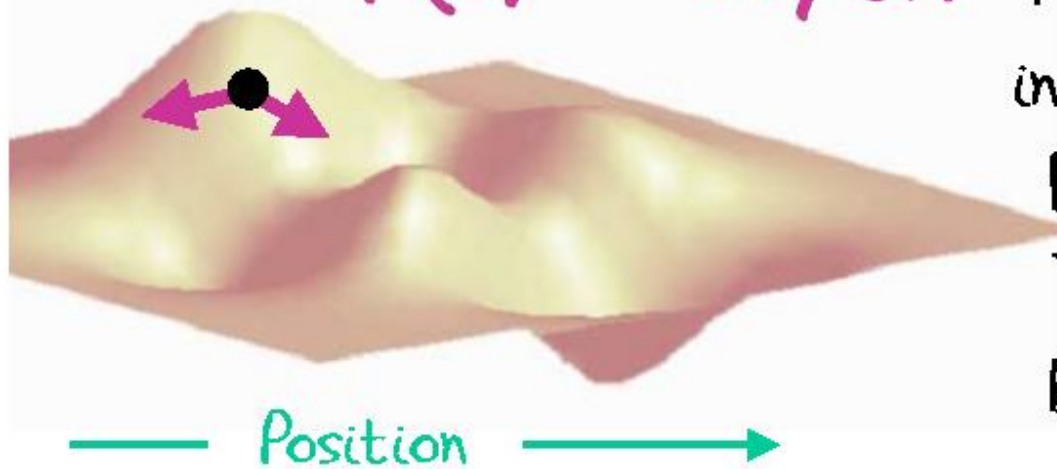


TOTAL POTENTIAL ENERGY. 2



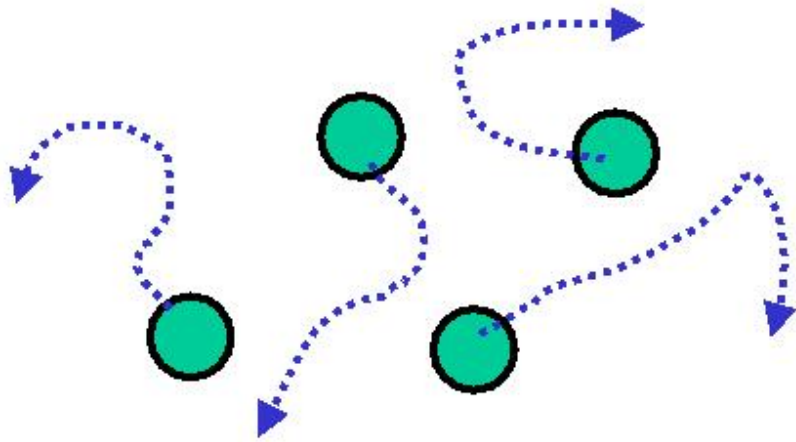
- The total potential energy or enthalpy fully defines the system, U .
- The forces are the gradients of the energy.

$$F(x) = -dU/dx$$



- The energy is a sum of independent terms for:
Bonds, Bond angles,
Torsion angles and non-bonded atom pairs.

MOLECULAR DYNAMICS THEORY



- Force = $-dU/dx$ (slope of potential, U); acceleration, $m a(t) = \text{Force}$.
- All atoms move together so force between atoms change with time.
- Analytical solution for $x(t)$ and $v(t)$ is impossible; numerical solution is trivial.

$$x(t+\Delta t) = x(t) + v(t)\Delta t + [4a(t) - a(t-\Delta t)]\Delta t^2/6$$

New position
Old position
Old velocity
Acceleration

$$v(t+\Delta t) = v(t) + [2a(t+\Delta t) + 5a(t) - a(t-\Delta t)]\Delta t/6$$

New velocity
Old velocity
Acceleration

$$U_{\text{kinetic}} = \frac{1}{2} \sum m_i v_i(t)^2 = \frac{1}{2} n k_B T$$

Kinetic energy
Atomic masses, velocities
Number of coordinates (not atoms)
Boltzmann's Constant
Temperature

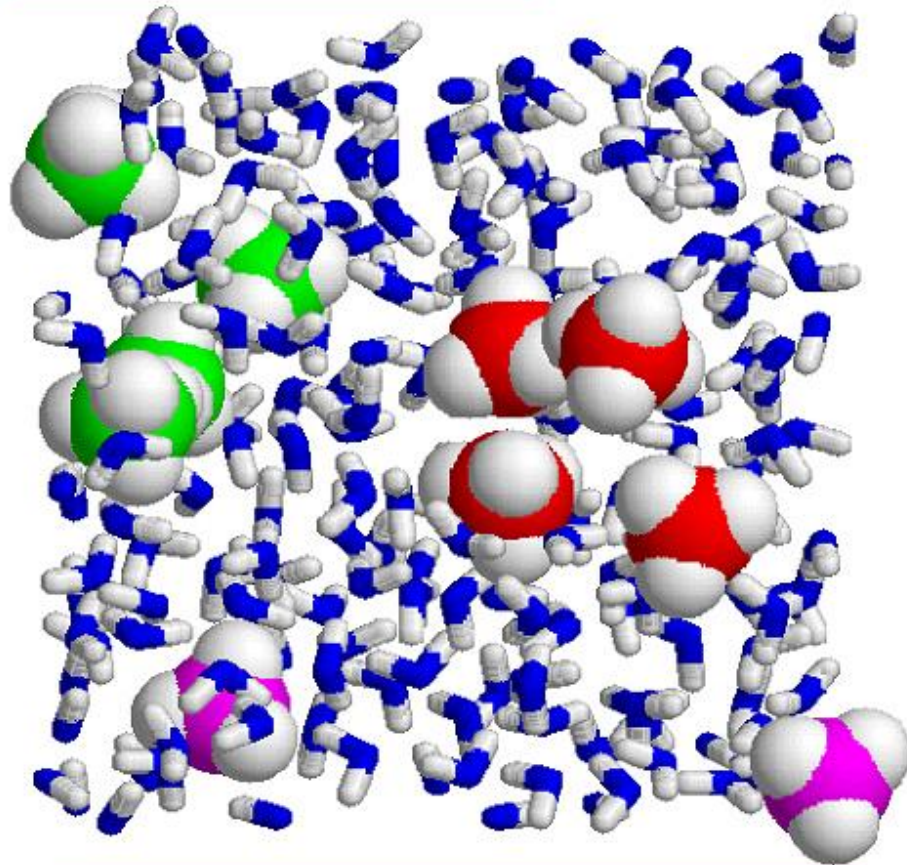
Time step, Δt , must be very small at 10^{-15} seconds or 0.001 ps.

Total energy ($U_{\text{potential}} + U_{\text{kinetic}}$) must not change with time

HYDROPHOBIC EFFECT

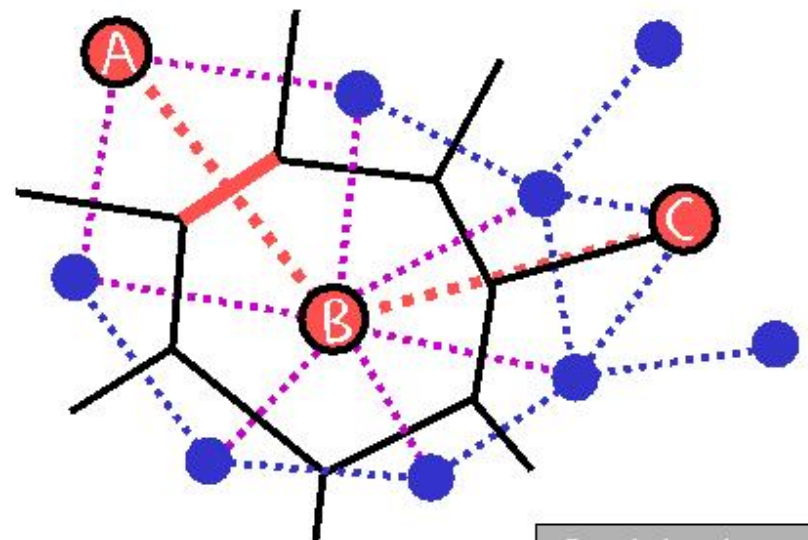
SIMULATING THE HYDROPHOBIC EFFECT

Tanya Raschke



Box with periodic boundaries.

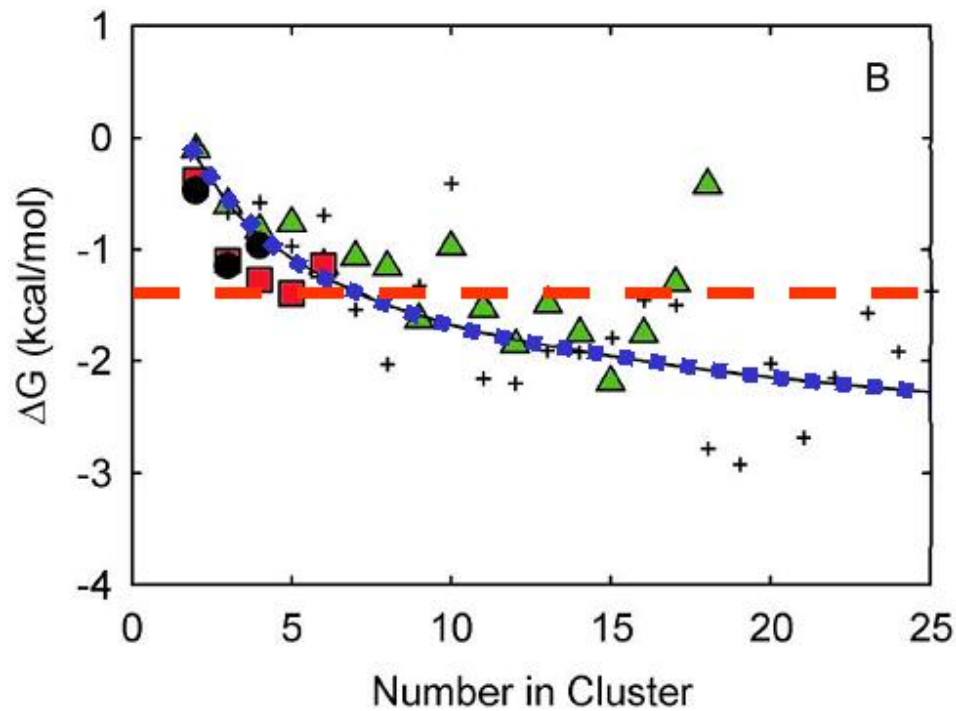
- 1 nanosecond MD simulations in periodic water boxes with from 30mM to 3 Molar hydrocarbon solution. Encad with F3C water (1996).
- Measure cluster formation by Voronoi. $d(AB) = d(BC)$, but only A, B touch.



MOVIE OF BENZENE
MOLECULAR DYNAMICS IN
WATER AT ROOM
TEMPERATURE



HYDROPHOBIC ENERGY IS COOPERATIVE



- $\Delta G_N = -kT \log[C_N / (C_{N-1} C_1)]$

- Assume clusters are close-packed spheres:

$$V_N = NV_1$$

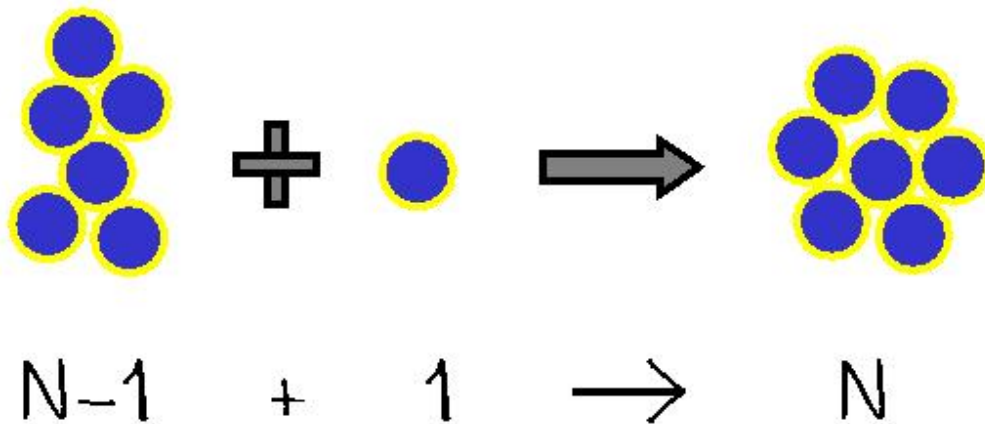
$$A_N = \alpha(V_N)^{2/3} = \beta(N)^{2/3}$$

$$\Delta A_N = \beta[(N)^{2/3} - (N-1)^{2/3}]$$

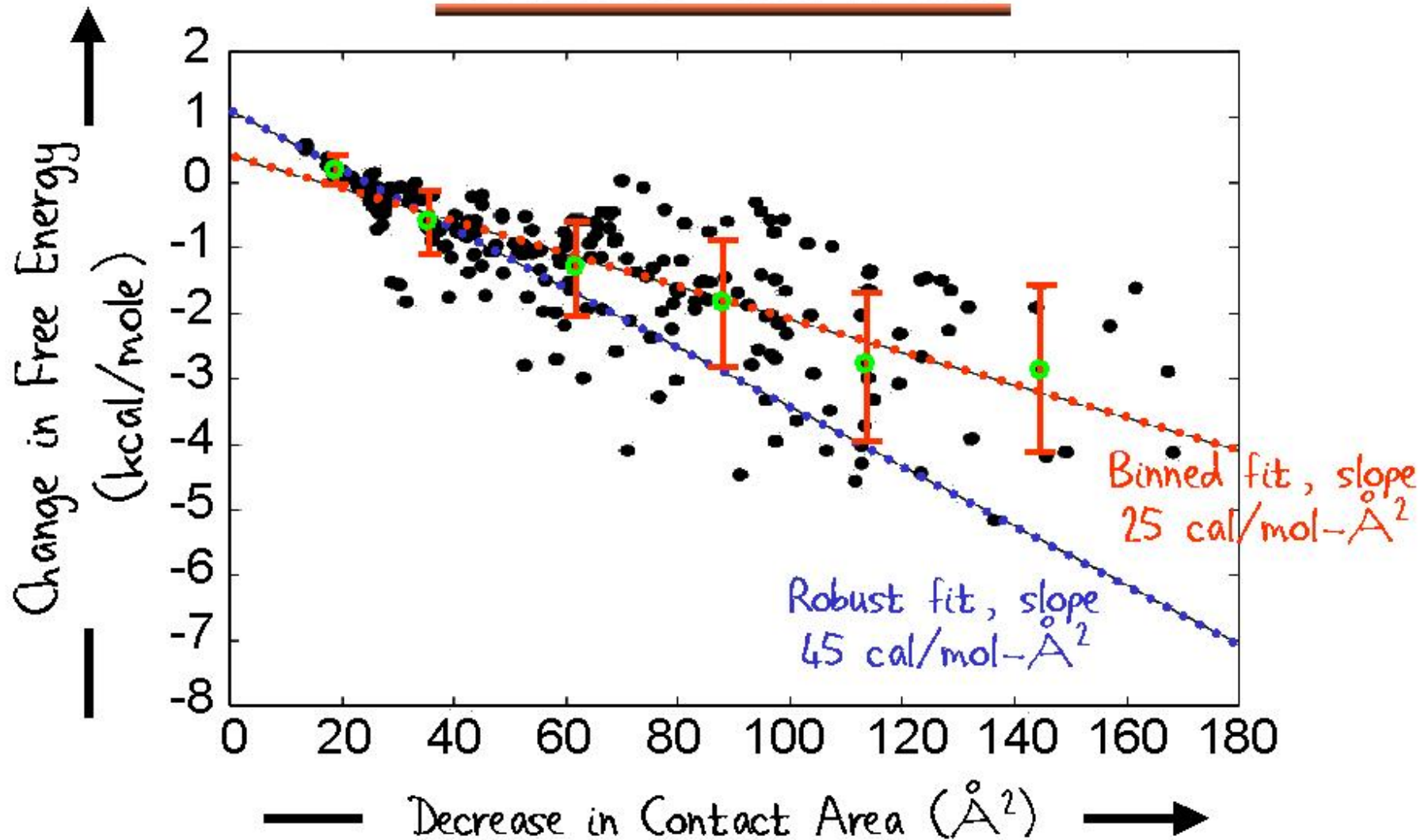
- If $\Delta G_N = \gamma \Delta A_N$, then

$$\Delta G_N = \gamma \beta [(N)^{2/3} - (N-1)^{2/3}]$$

- Determine γ by fitting with $[(N)^{2/3} - (N-1)^{2/3}]$.



HYDROPHOBIC ENERGY DEPENDS ON BURIED SURFACE

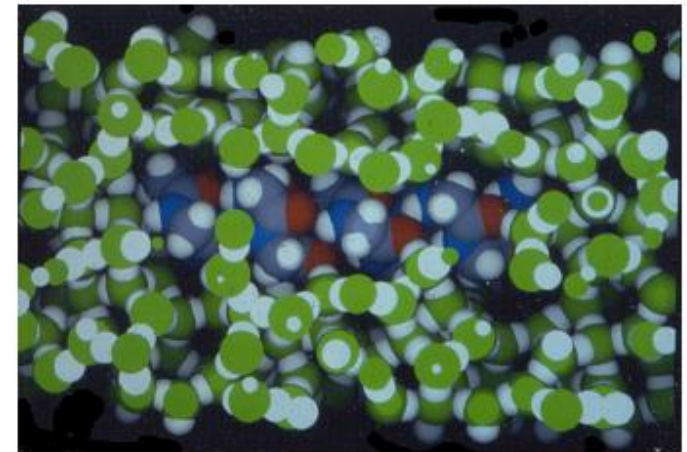
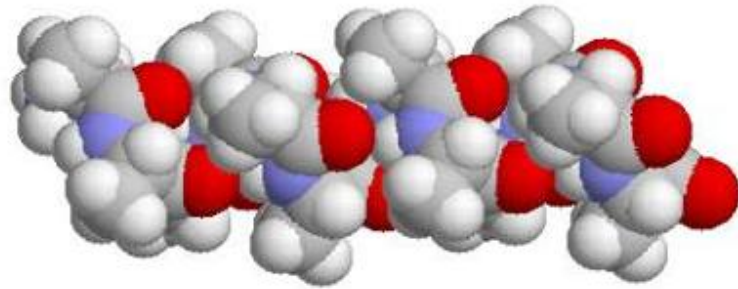


Constant of proportionality matches experiment.

SIMULATE UNFOLDING

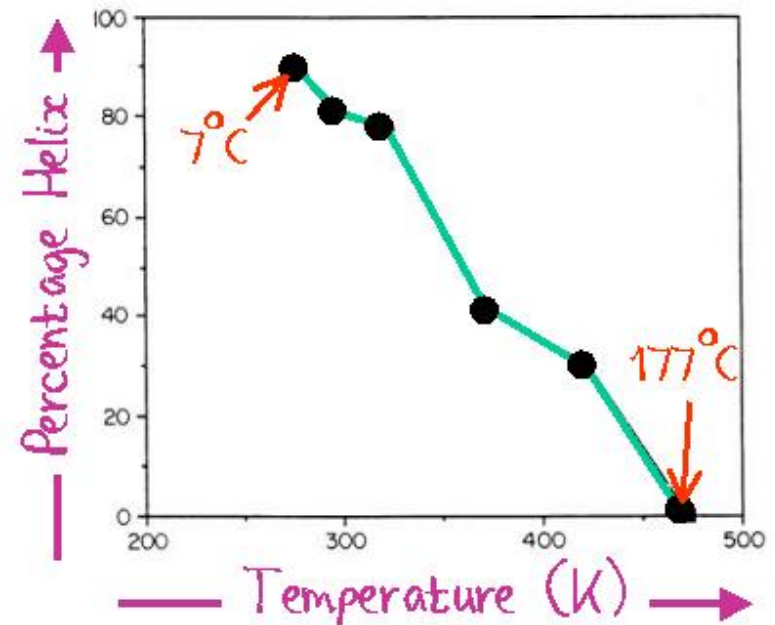
UNFOLD THE α -HELIX

13 Alanine residues

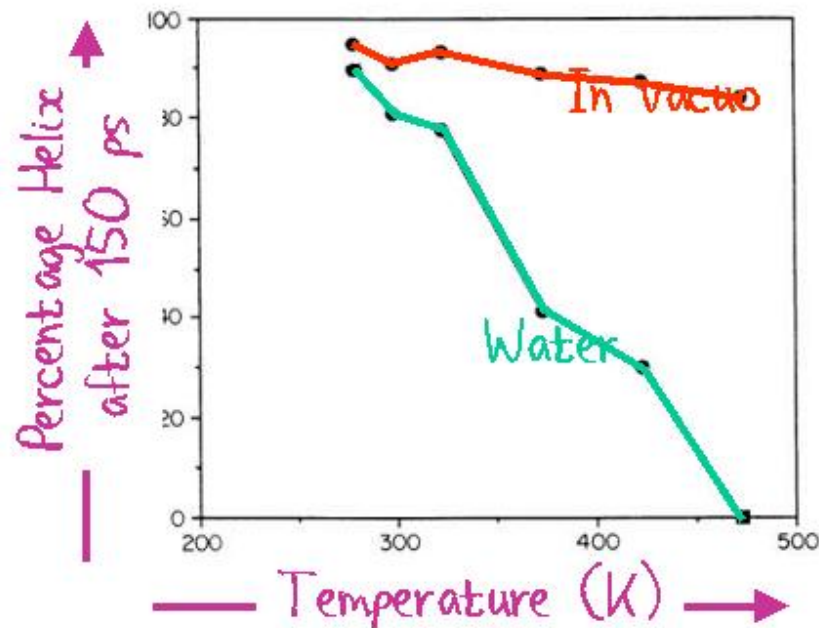


- Start as an ideal α -helix in a box of water.
- Run 200 ps (100,000 time steps) of molecular dynamics at six different temperatures.
- Record percentage α -helix formed for last 50 ps.
- See temperature-induced melting on picosecond time-scale.

Put it in a box of water.



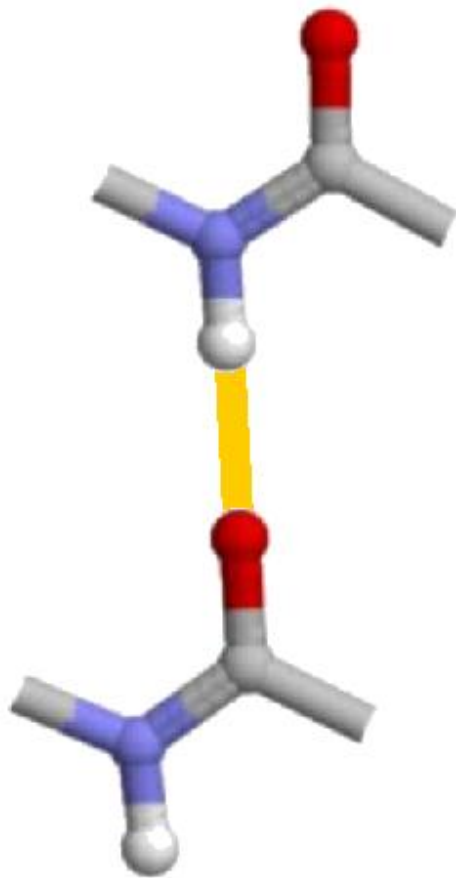
α -HELIX LESS STABLE IN WATER



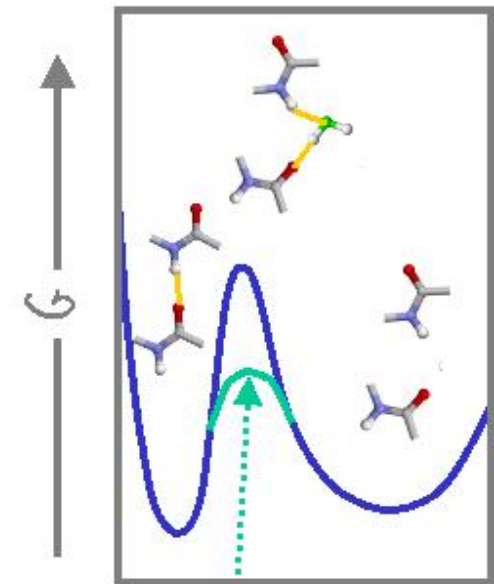
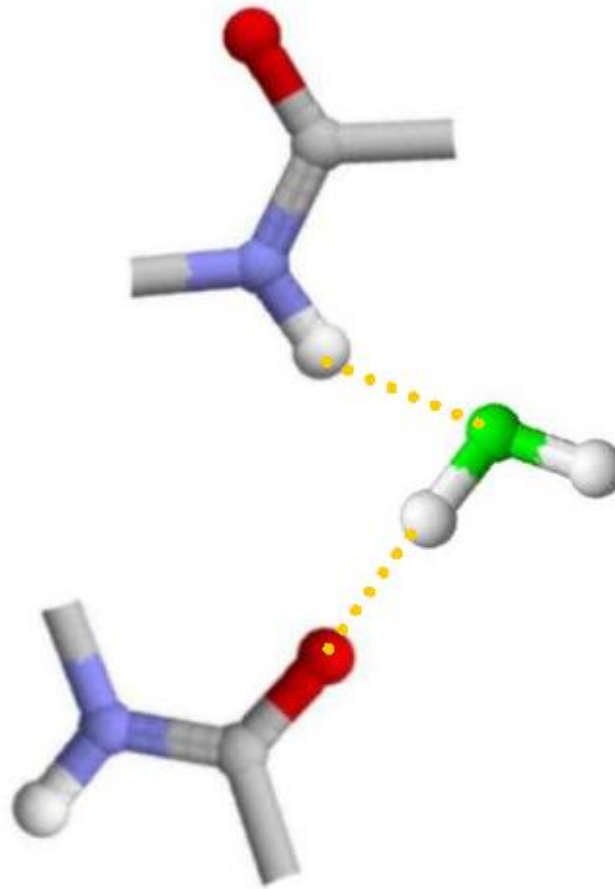
- In vacuo the helix is very stable even at high temperature.
- In water the helix is unstable at high temperature.
- The rate of melting depends on temperature.
- Happens because water molecules stabilize the transition state.

WATER ALLOWS HYDROGEN BONDS TO BREAK

Intact hydrogen bond in helix



Hydrogen bond is breaking

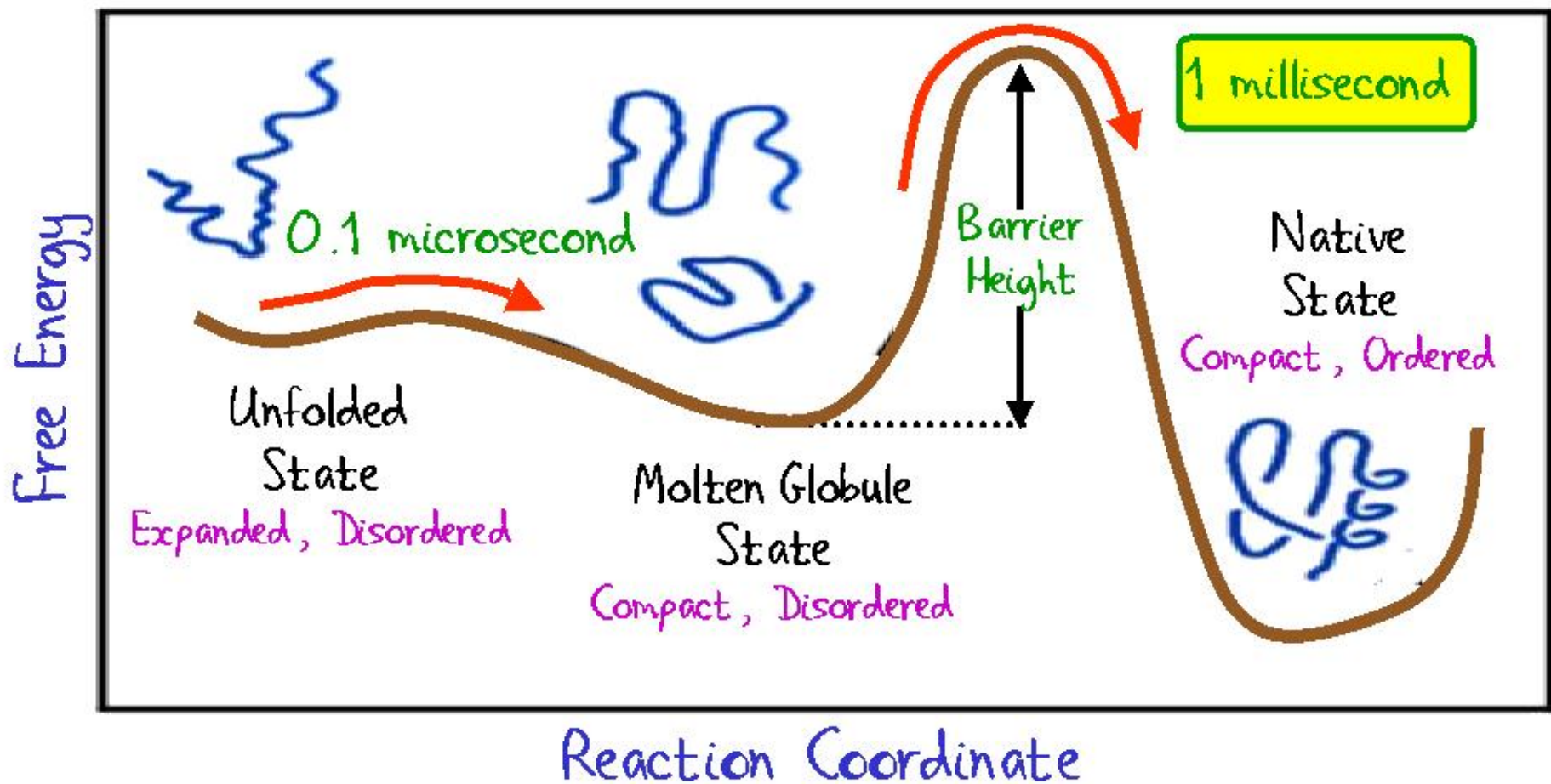


Free Energy barrier between states is much lower in water

- Water catalyzes the breakage of hydrogen bonds by stabilizing the transition state.

SIMULATE FOLDING

SIMULATING FOLDING IS DIFFICULT?



- Simulation of 1 millisecond requires 10,000 CPU years!
- Must get over high barriers & many degrees of freedom.

MASSIVE COMPUTATIONAL RESOURCES

- Empty Supercomputers.
- Blue Gene (IBM).
- Folding@home (Vijay Pande).

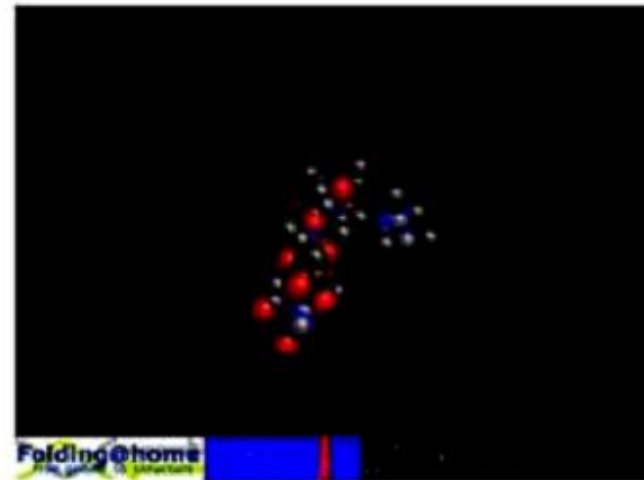
FOLDING@HOME



<http://www.stanford.edu/group/pandegroup/Cosm/>

Using Folding@home

- [Project Goals: solving the protein folding problem](#)
 - [How you can help](#)
 - [Download \(New!! Version 1.33\)](#)
 - [How to install our software](#)
 - [Frequently asked questions \(FAQ\)](#)
 - [Contact Folding@home \(Help Center\)](#)
 - [Folding@home discussion board](#)
- Fold proteins on 20,000 computers using the program as a Screen Saver!



Join Folding@home by running our [screen saver](#) or client software

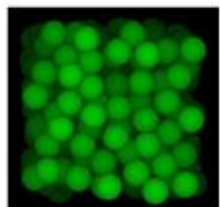
Like SETI@home

PANDE MOVIE
 α -HELIX FOLDING

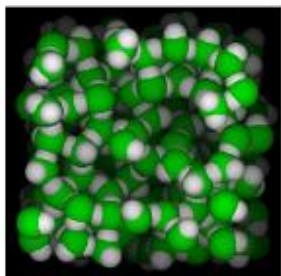


50 YEARS OF SIMULATION

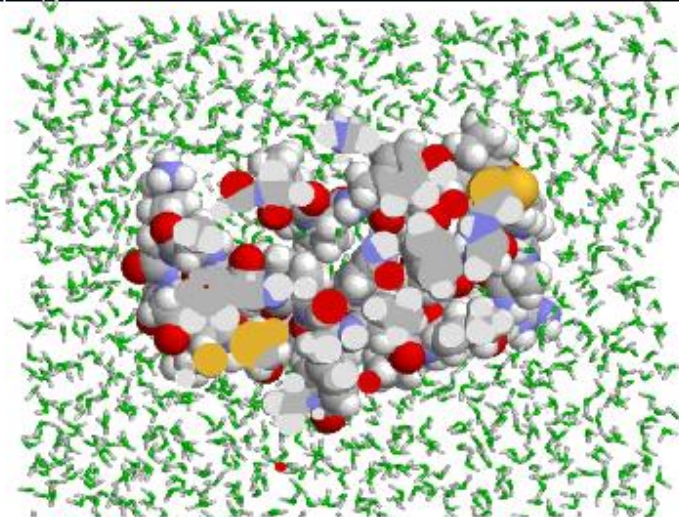
- We have 10,000,000 times more resources.
- Systems have become larger (100 times).
- Runs have become longer (100,000 times).
- Energy functions have become simpler.
- Fit reality well. **Nothing bad has happened!**



1955 Argon



1970 Water



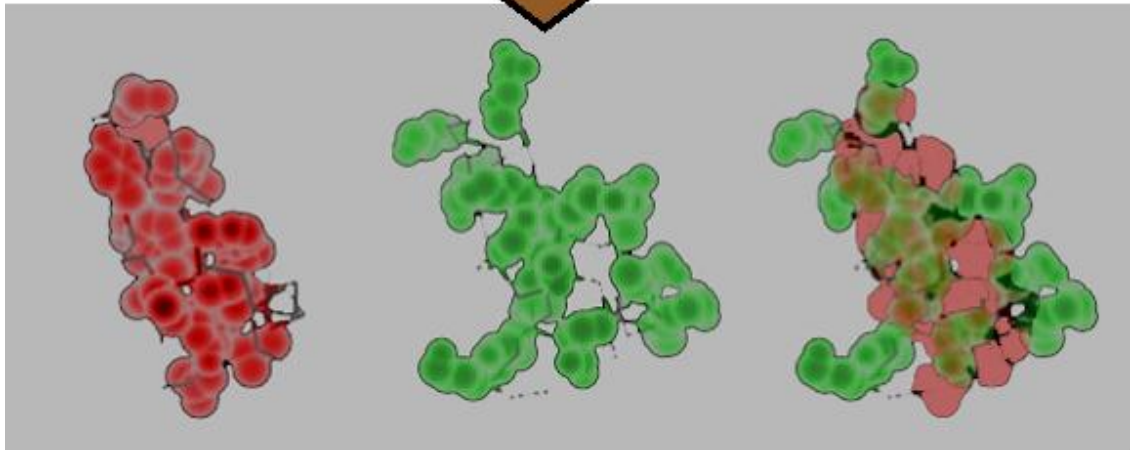
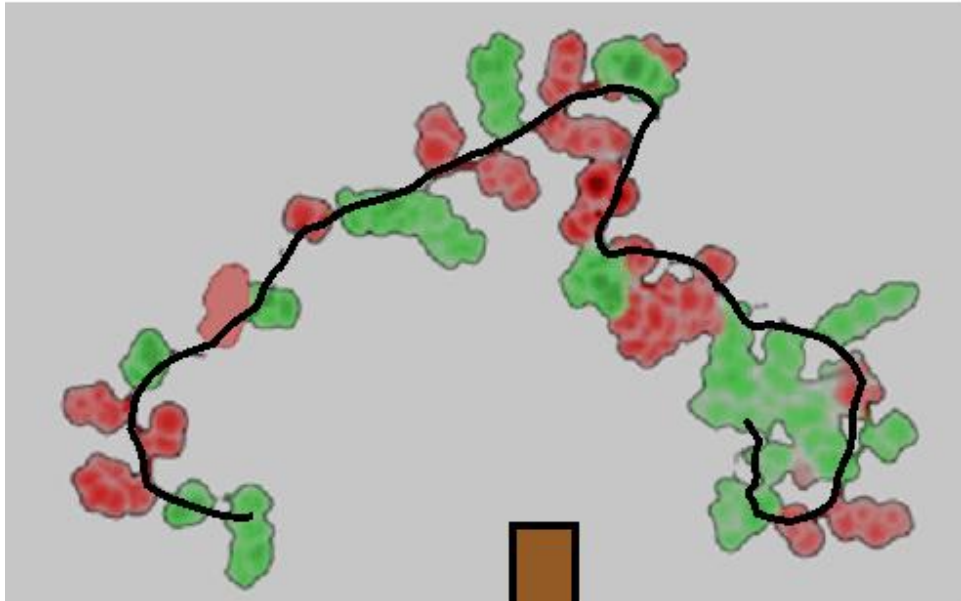
1988 Protein in Water

?

2005

PREDICTION

WHAT DRIVES FOLDING?



Hydrophobic

Hydrophilic

All Residues

- Protein is a chain.
- Self-avoiding and close packed.
- Residue preferences:
 - Inside/Outside
 - Specific Neighbors

Red are hydrophobic, like to be away from water

Green are hydrophilic, like contact with water

Discrimination Paradigm

A PARADIGM FOR PREDICTING STRUCTURE

DECOYS

- Construct a large number of possible folded shapes (Decoys).

DISCRIMINATION

- Select the correct, native fold.



Need a good energy function

THE CASP EFFECT

- Critical Assessment of Structure Prediction.
- Predict what no one knows. John Moult
- Predict what is about to be known.
- Carefully control evaluation and assessment. Competition?
- Meet to discuss what went wrong and what went right.
- Have had CASP1 ('94) through CASP4 ('00).

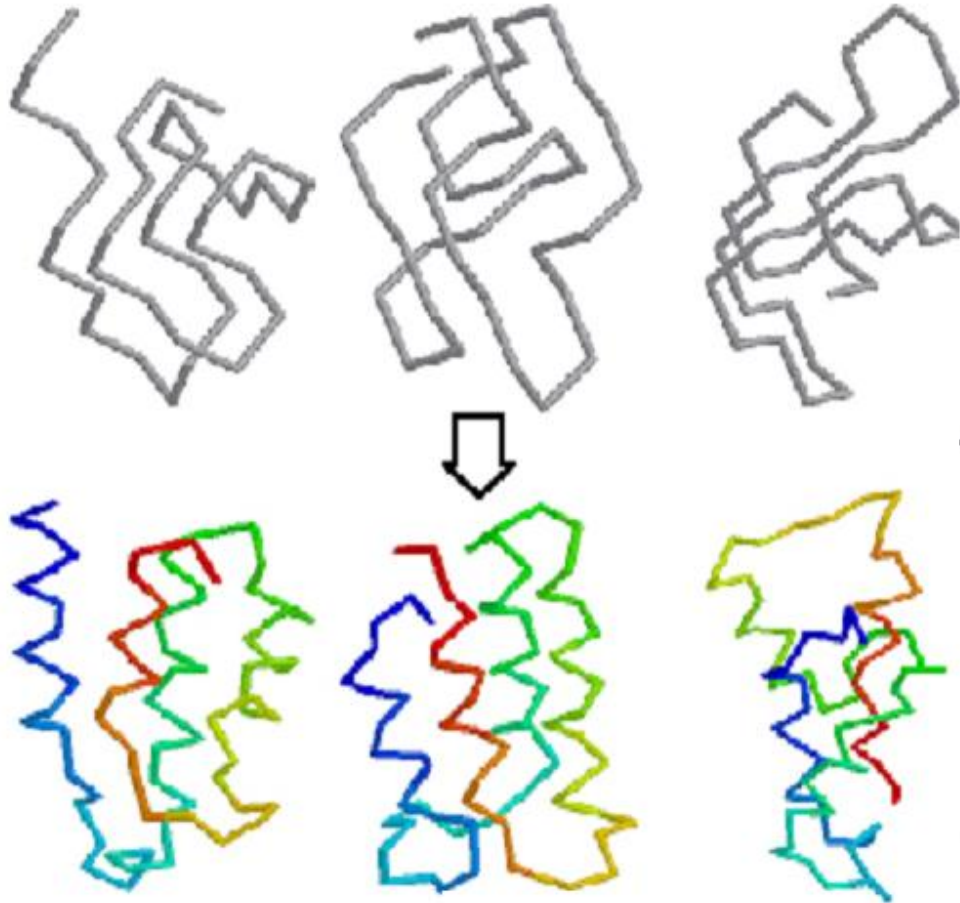
Moult, Pedersen, Judson and Fidelis, A Large-Scale Experiment to Assess Protein Structure Prediction Methods. *Proteins*. 23: ii-v (1995)

HIEARCHICAL REDICTION

1998

HIERARCHICAL STRUCTURE PREDICTION

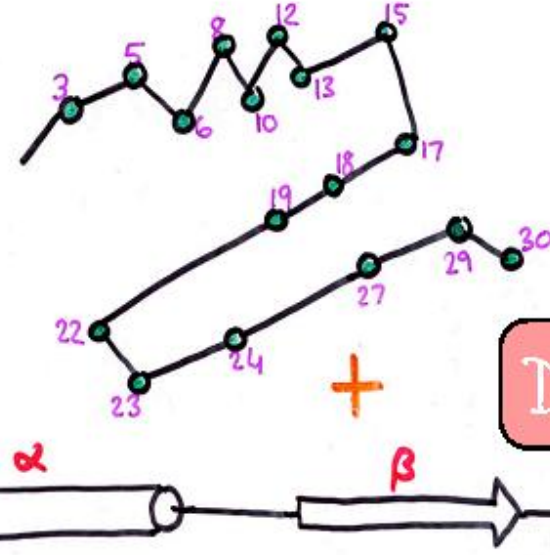
Use a $\sqrt{3}$ -state lattice model.



Use a 4-state off-lattice model.

Yu Xia

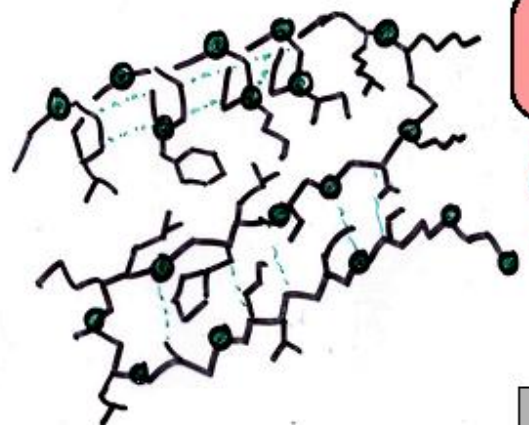
Ram Samudrala



1 of 10,000 low-energy shapes.

David Hinds

Predicted secondary structure



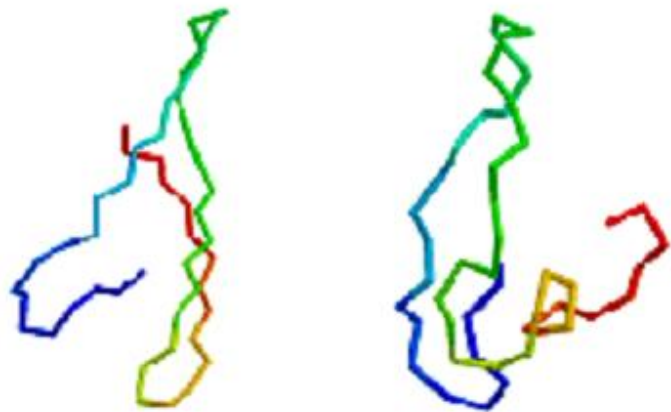
Britt Park

Add all atoms in full detail.

HIERARCHICAL PREDICTION DOES WELL

Does well at CASP3 (Critical Assessment of Structure Prediction, Asilomar 1998).

T46/adg 7.5 Å (49 residues; 66113)



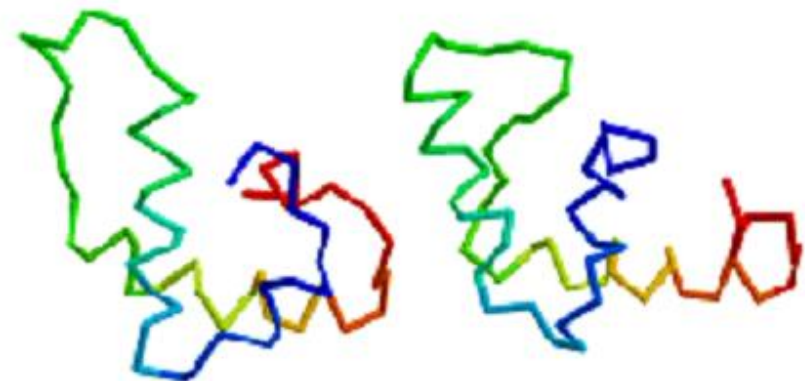
* T56/dnab 6.8 Å (60 residues; 67126)



** T59/smd3 6.7 Å (46 residues; 3075)



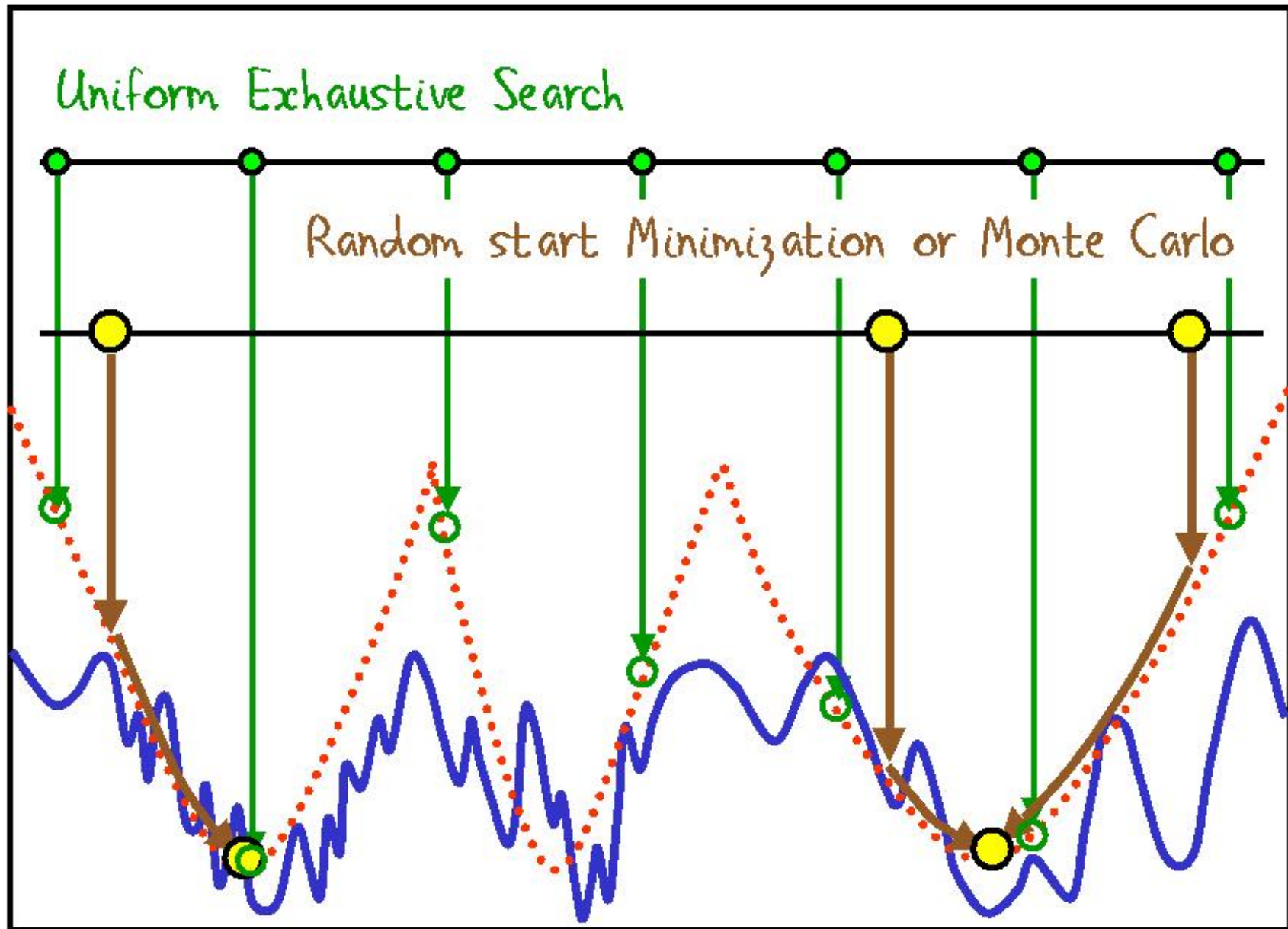
** T61/hdea 7.4 Å (66 residues; 974)



Samudrala, Xia, Huang & Levitt. Bona Fide Ab Initio Protein Structure Prediction Using a Combined Hierarchical Approach. *Proteins*, 37 (3S): 194-198 (1999).

SPECIAL POTENTIALS 2000

SAMPLING ANT LION TOWN POTENTIALS



Energy Minimization

ALL-ATOM ENERGY MINIMIZATION

- Minimize all-atom energy with respect to all torsion angles.
- Augment the normal potential energy function with:
 - Cooperative hydrogen bonds.
 - Cooperative hydrophobic interactions.
 - Forced exposure of charges.

Chen Keasar

POTENTIAL ENERGY IN TORSION SPACE

$$U = \sum K_{\phi} [1 - \cos(n\phi + \delta)]$$

All Torsion Angles

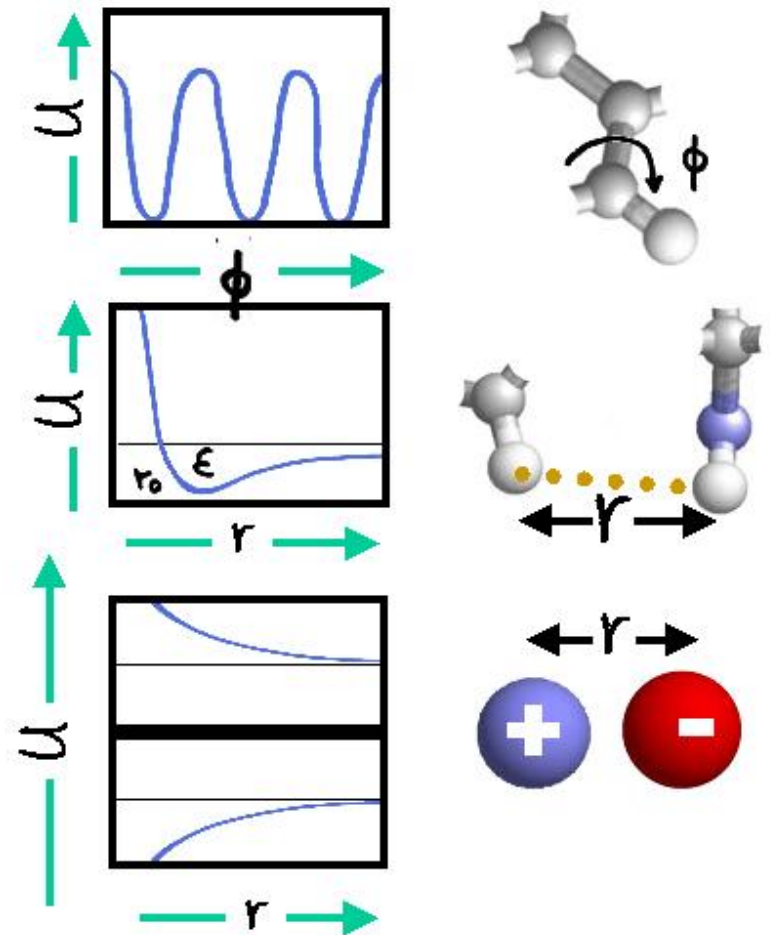
$$+ \sum \epsilon \left[\left(\frac{r_0}{r} \right)^{12} - 2 \left(\frac{r_0}{r} \right)^6 \right]$$

All nonbonded pairs

$$+ \sum \frac{332 q_i q_j}{r}$$

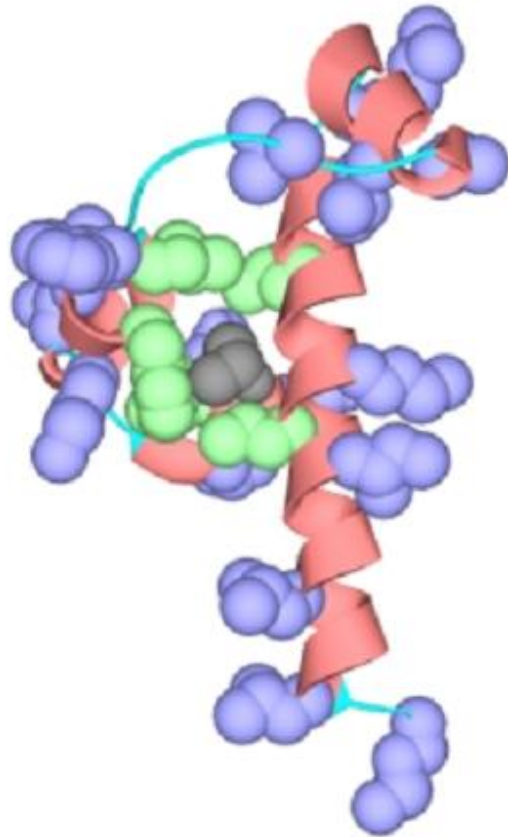
All partial charges

- A protein with N residues has about $4N$ (ϕ, ψ, χ) single bond torsion angles. The same protein has about $50N$ Cartesian coordinates (x, y, z).

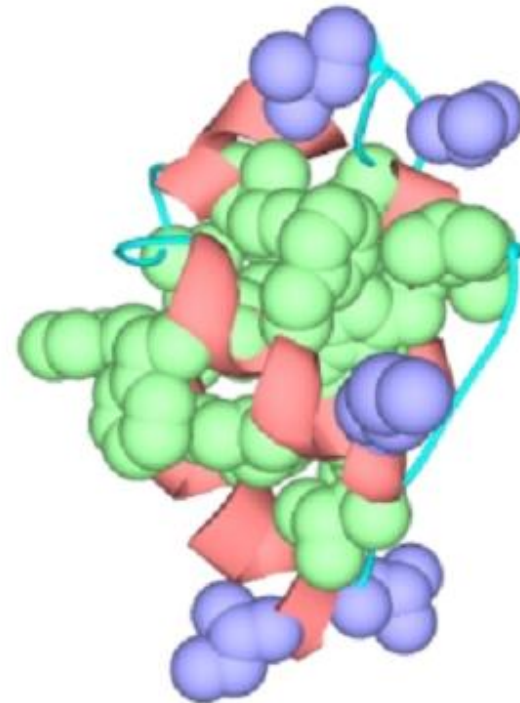


COOPERATIVE HYDROPHOBIC PACKING

- Cooperative hydrophobic compaction makes a good core.



Original Potential

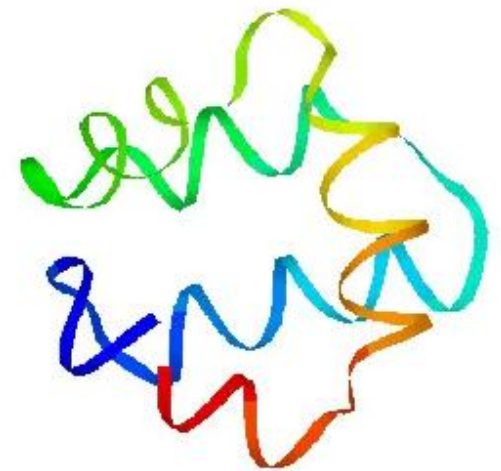


Modified Potential

STRUCTURE PREDICTION BY MINIMIZATION

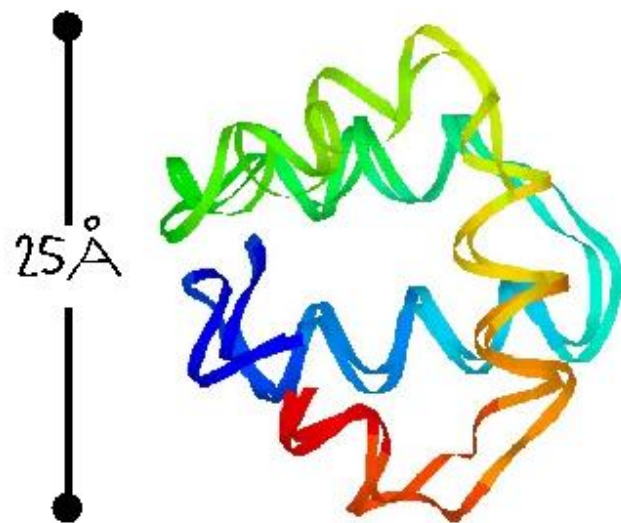
- Minimize special energy function with respect to torsion angles (ϕ, ψ, χ).
- Add energy terms for cooperative hydrogen bonds and hydrophobic compaction.

$10\text{\AA} = 1\text{ nm}$

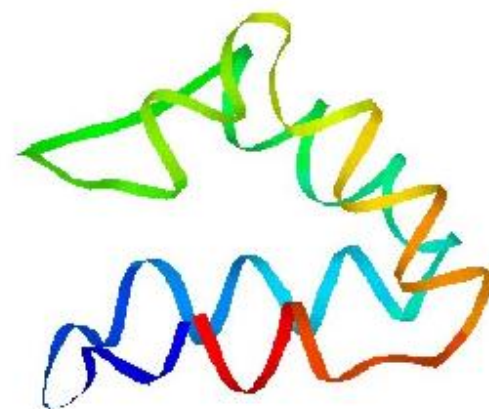


Structure of T102

This method did well at
CASP4, 2000.



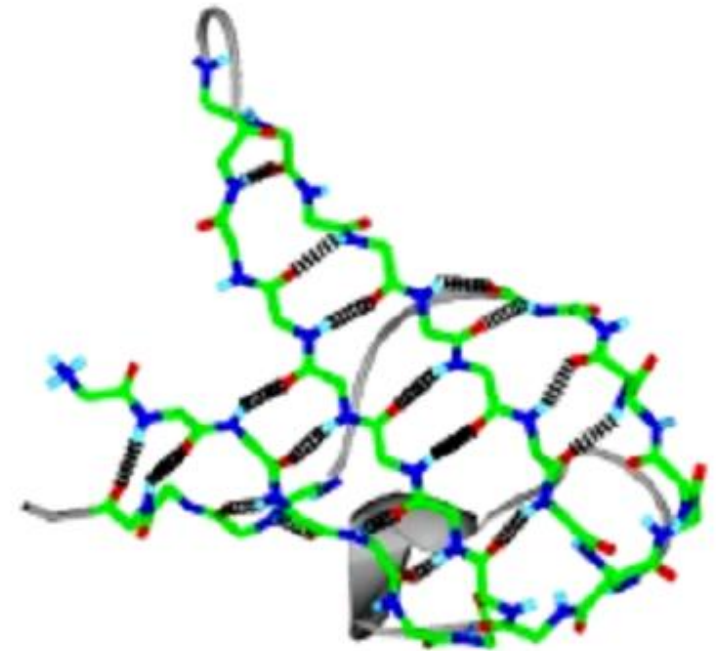
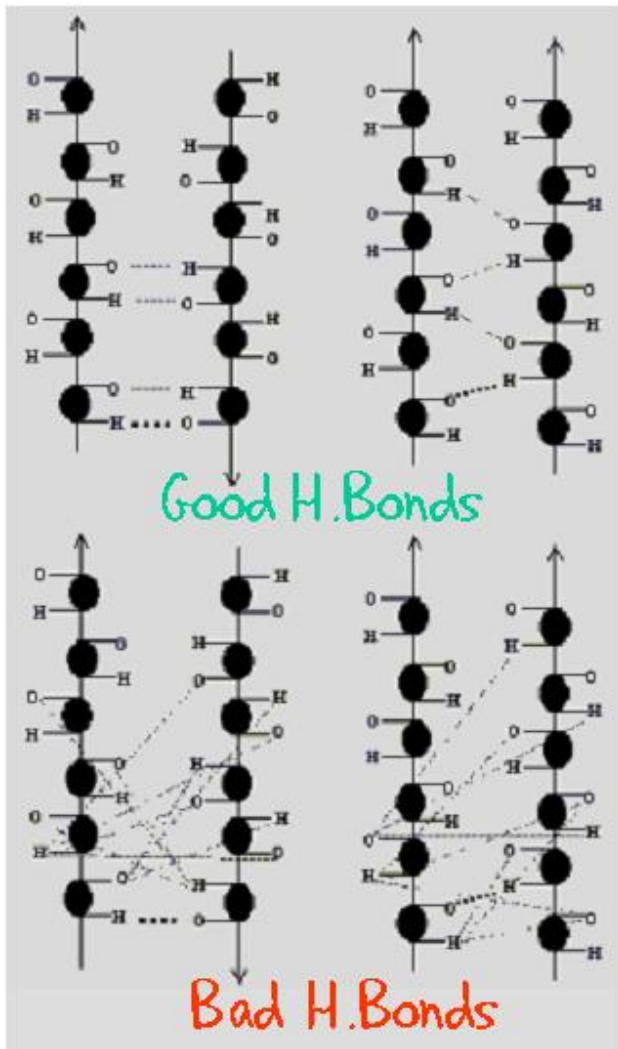
Best T102. RMS = 3.3\AA



T102 Submitted. RMS = 5.0\AA

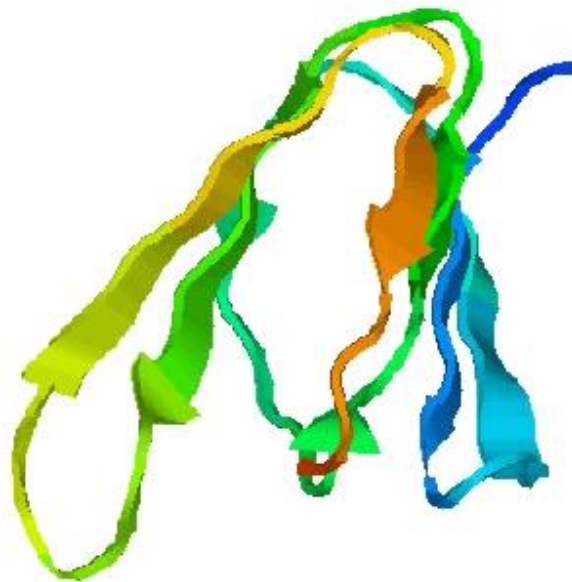
COOPERATIVE HYDROGEN BONDS

- Cooperative hydrogen bonds give rise to good secondary structure.

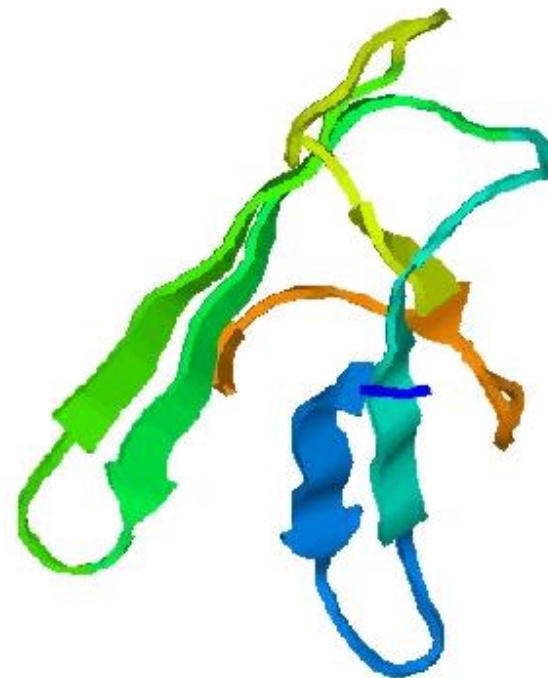


ALL- β PREDICTION SUCCESS

- All- β sheet proteins are the hardest to predict.
- Torsion minimization does well on T114, an all β -protein.



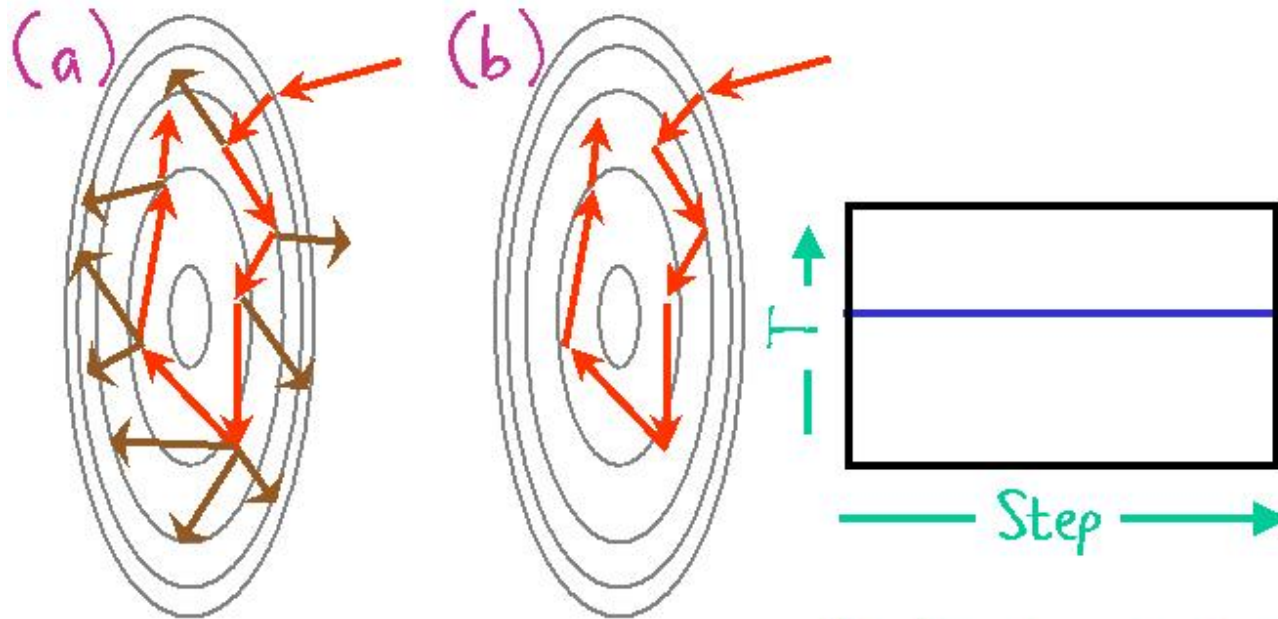
Native Structure.



Prediction is somewhat similar.

Segment Monte Carlo

MONTE CARLO METHODS



• Normal Monte Carlo:

Make random moves and accept some of them (Metropolis).

• Simulated Annealing:

Reduce T , the temperature, as the run proceeds.

FRAGMENT MONTE CARLO

Fragment Library



RVL



RFL



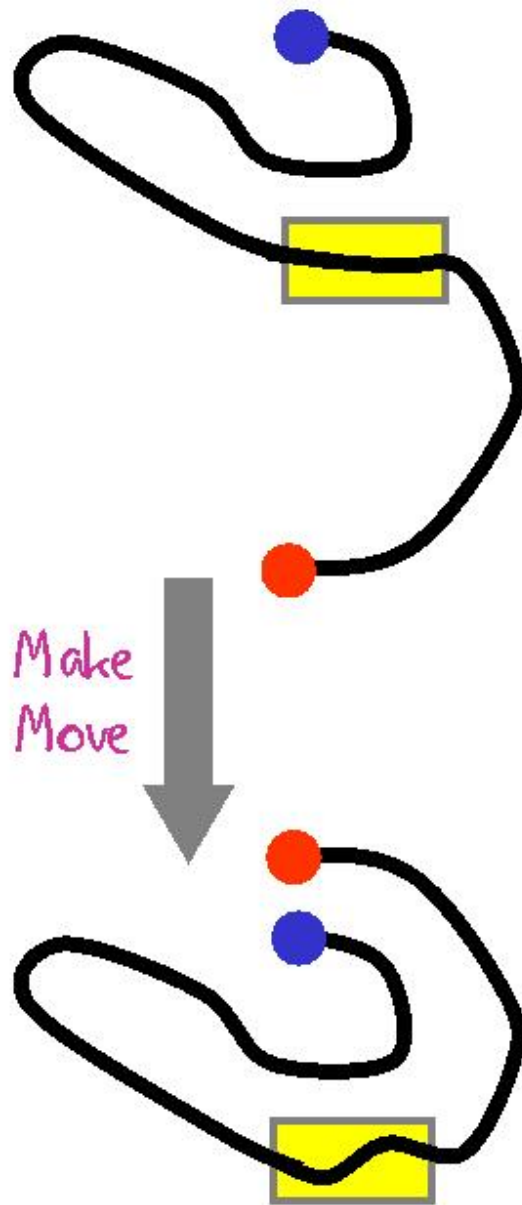
EVL



KVI

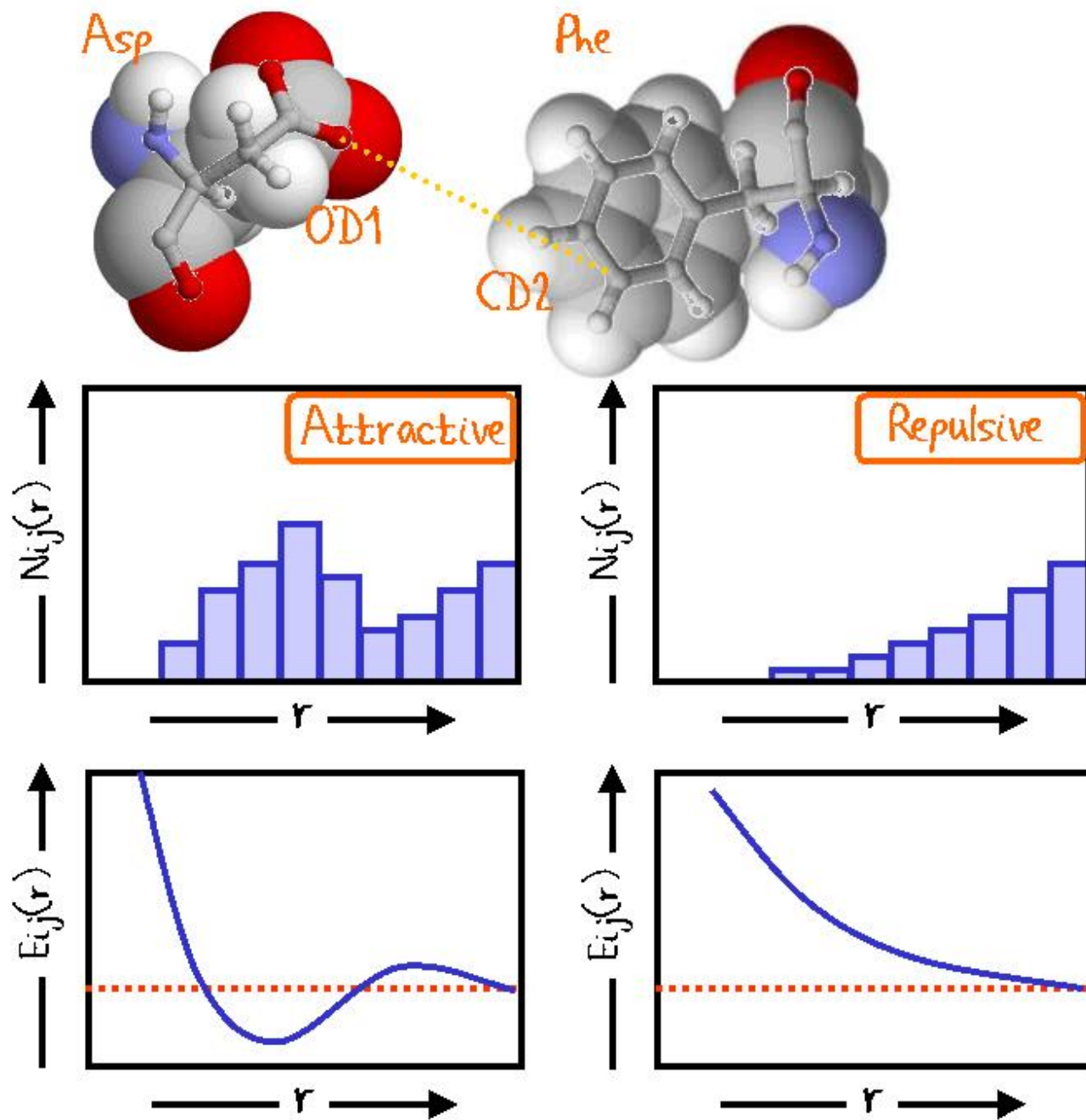


KVY



- Make a library of small fragments of similar sequence.
- Swap in a new fragment by setting six (φ, ψ) torsion angles.
- Accept move by Monte Carlo and anneal.

KNOWLEDGE-BASED ENERGIES



- Get distribution of distances between pairs of atom centers of a particular type, e.g. D-OD1...F-CD2.

- Normalize and take log to get Energy score:

$$E_{ij}(r) = \log[N_{ij}(r) / M_{ij}(r)]$$

SEGMENT FOLDING PREDICTION

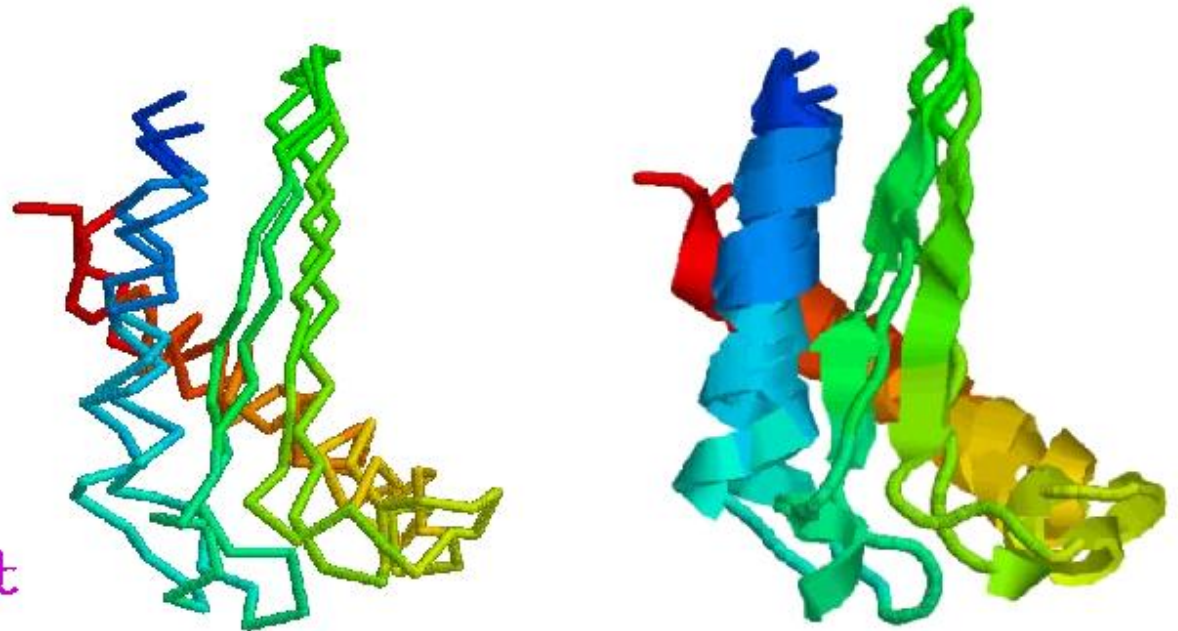
- Do Monte Carlo moves with respect to (ϕ , ψ) torsion angles. Simulated annealing.

Ram Samudrala

- Use all-atom Knowledge-Based energy function.
Add terms to enforce compaction.

- Get reasonable (ϕ , ψ) angles from real protein fragments.

- This method does well at CASP4, Asilomar '00.



T0110 Fit 80 residues to 4.0\AA

HARD PROBLEMS

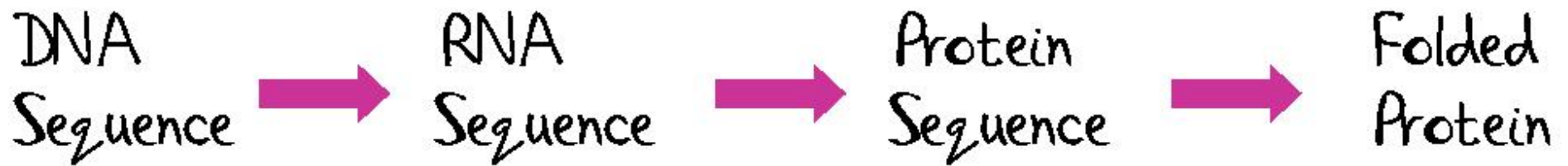
WHY IS FOLDING SO HARD?

- Many different specific interactions.
- Cooperativity of the underlying interactions.
- Three-dimensional with very many possible spatial arrangements.
- Violates Crick's Law of Hard Problems.

WHY ARE WE GETTING BETTER AT FOLDING?

- Peer pressure (CASP)?
- Faster computers?
- Many more sequences?
- More structures?

INFORMATION + PHYSICS = LIFE



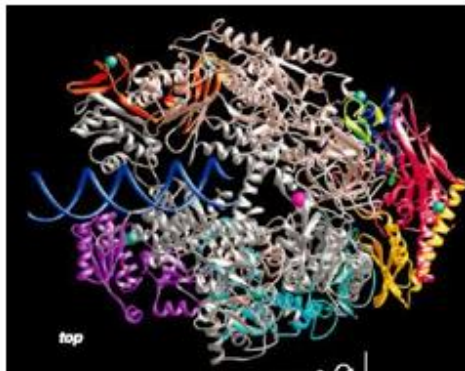
• in silico

Easy:
Change
T to U

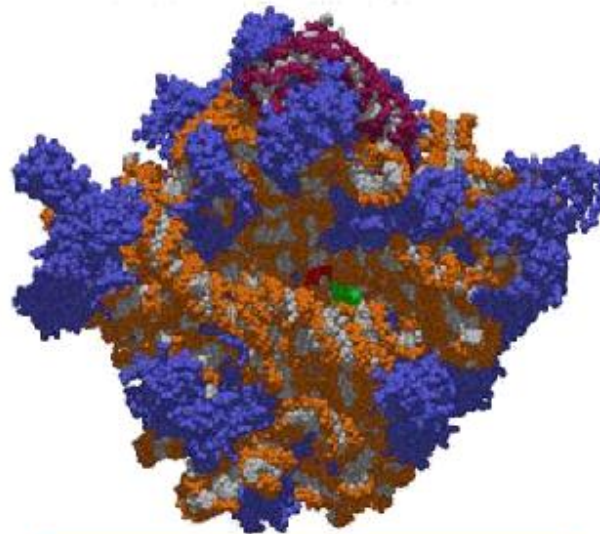
Easy:
Triplet
Code

Hard: Folding is
many body
simulation

• in vivo



Hard: Transcription
Polymerase



Hard: Translation
Ribosome

Easy: Folding is
free by laws of
physics

ACKNOWLEDGEMENTS

PEOPLE

- Tanya Raschke
- Erik Sandelin
- Boris Fain
- Patrice Koehl
- Michael Sykes
- Yu Xia
- Rachel Kolodny
- Chen Keasar
- Nizar Batada
- Chris Summa

SUPPORT

- NIH (NIGMS); DOE (SBI); NSF (ITR)

WEB

- <http://csb.stanford.edu/levitt>
- <http://biospace.stanford.edu>
- <http://dd.stanford.edu>
- <http://astral.stanford.edu>

- Papers
- Lecture
- Thesis

THE END

HISTORICAL RECORD OF BEST PREDICTIONS AT CASP

CASP & YEAR	NUMBER TARGETS	BEST RESULT <Q3>	RESULT GROUP
CASP1 1994	6	63	Rost & Sander
CASP2 1996	24	70	Rost
CASP3 1998	18	75	Jones
CASP4 2000	28	80	Jones

- Steady improvement of about 5% per CASP (every two years)

NOTES

- PDF files on home page
- Searching www.google.com

