

Monte Carlo Simulation of Protein Folding

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PHYS 2600 Computational Physics Final presentation

Introduction

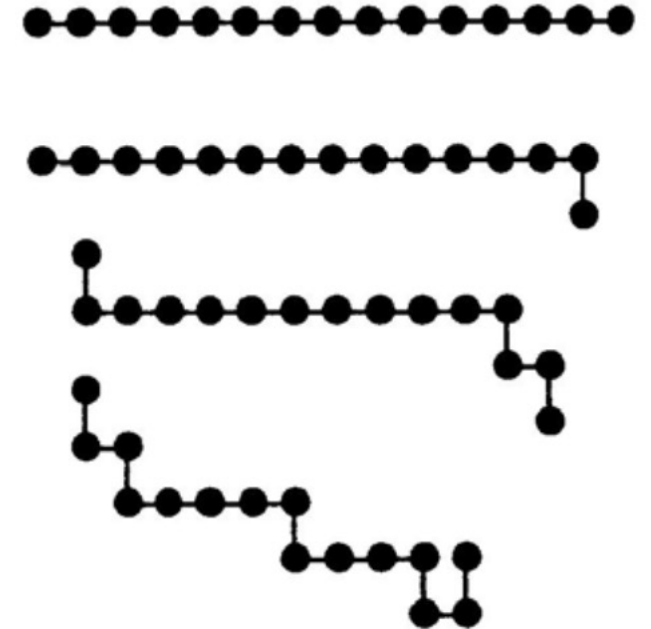
- Protein folding is one of the most important topics in biology and biophysics.
- When a protein works in a biological activity, its structure is folded, aka **folded state**
- Heating or some chemical environments can cause protein denaturation and unfolding. *[1]*
- Monte Carlo method can give us a rough result quickly and help us understand this system.
- This project: simulate the process of protein folding and study the effect of temperature i.e. the denaturation process

Model

- Assume the protein is an amino acid chain consisting of **N** amino acids and located on a 2D grid. Each amino acid is on a grid point, forming a **self-avoiding walk** pattern.
- Neighboring amino acids are linked by covalent bonds.
- The total interaction energy of the protein:**[2]**

$$E = \sum_{\langle m,n \rangle} J_{A(m),A(n)}$$

$\langle m,n \rangle$ means a pair of **non-covalent neighboring** amino acids. $A(m)$ represents the type of m-th amino acid. J is the interaction coefficient between two amino acids.



*Figure 1: Four examples of possible states for an amino acid chain**[2]**.*

Algorithm

- Metropolis[3][5] and simulated annealing algorithm[4] is applied.
- In high T:
 1. Choose a node randomly: (x_{old}, y_{old})
 2. Choose one of its **next nearest neighbors** randomly: (x_{new}, y_{new})
 3. Test SAW.
 4. Calculate the change in energy ΔE after the movement.
 5. The acceptance probability is $e^{-\beta\Delta E}$, $\beta = \frac{1}{k_B T}$
- In low T: Set high T initially, then decrease T gradually

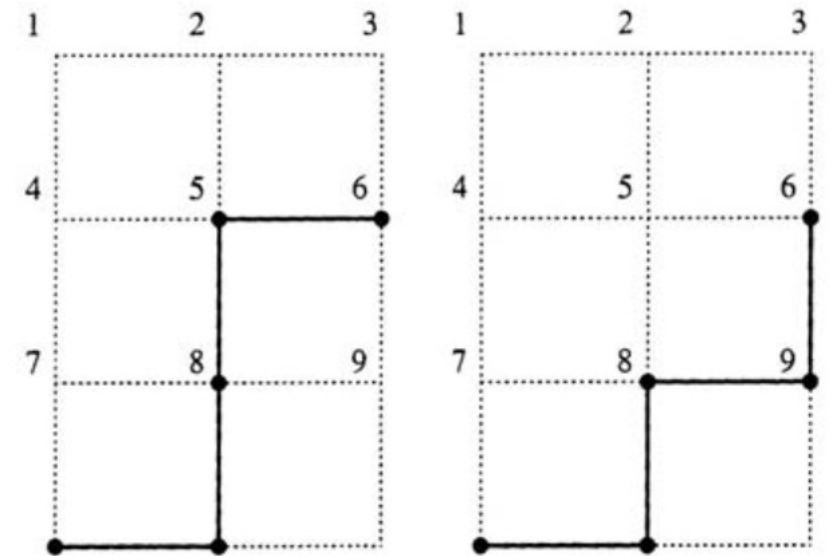


Figure 2: An example for an iteration[2].

Simulation Results: Equilibrium

100 amino acids, $T = 1.0$, with SA

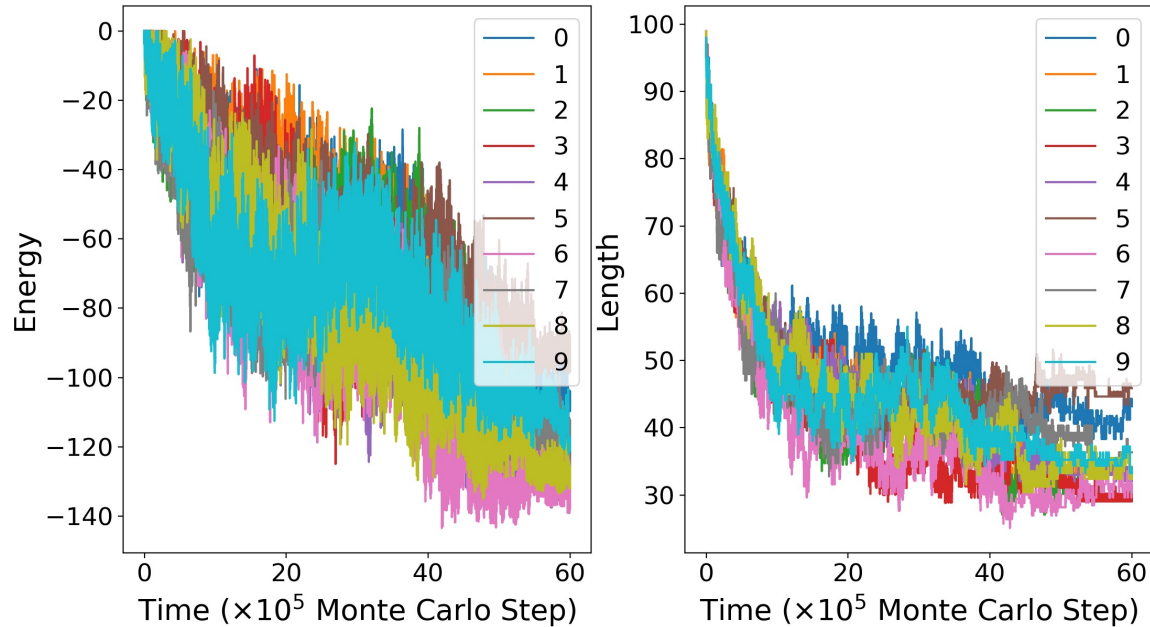


Figure 3: Evolution of the energy and **end-to-end distance** (length) of the protein consisting of 100 amino acids when $T = 1$. Simulated annealing method is applied in this simulation.

100 amino acids, $T = 1.0$, w/o SA

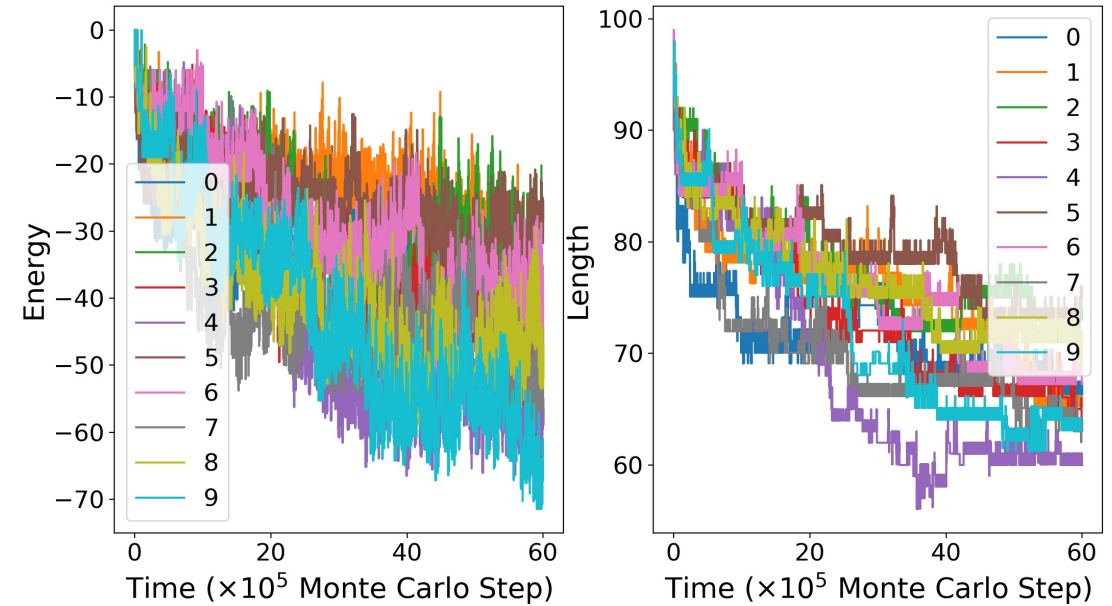


Figure 4: Evolution of the energy and end-to-end distance (length) of the protein consisting of 100 amino acids when $T = 1$.

- Simulated annealing method does work in low temperature.
- Simulating the system for many times is needed to improve statistical accuracy

Simulation Results: Effect of T & J

15 amino acids

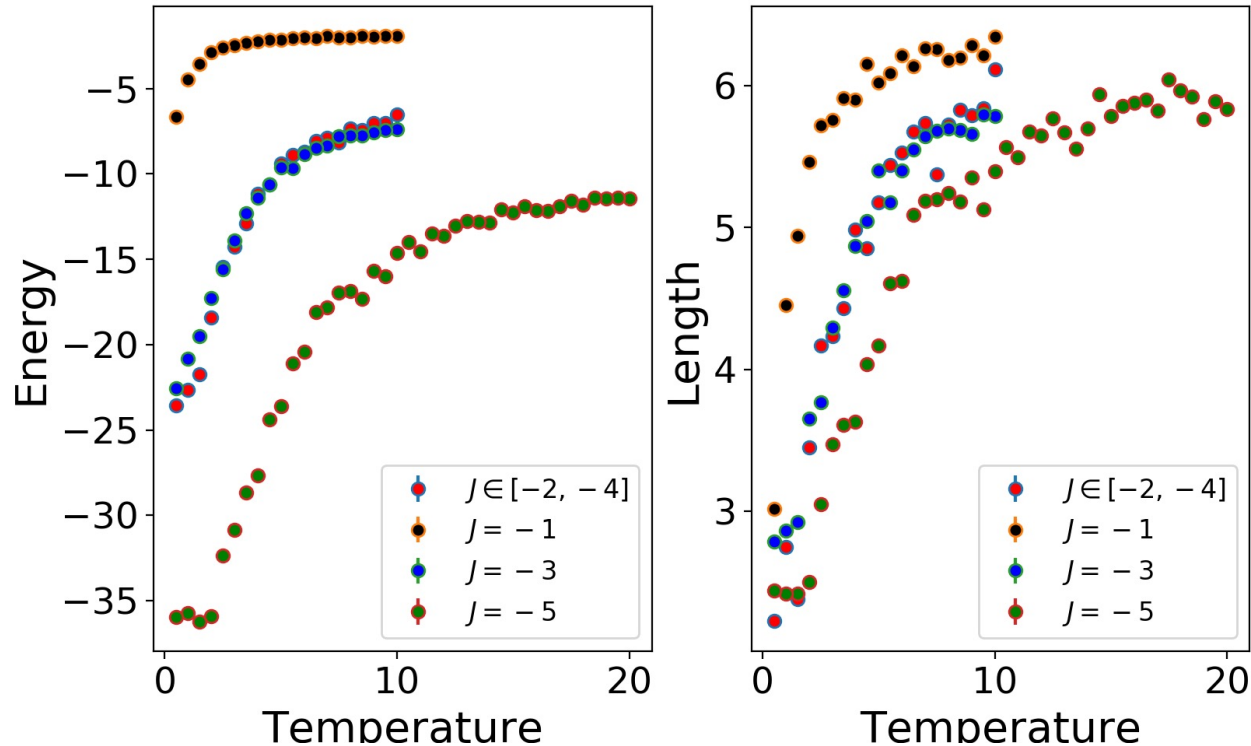
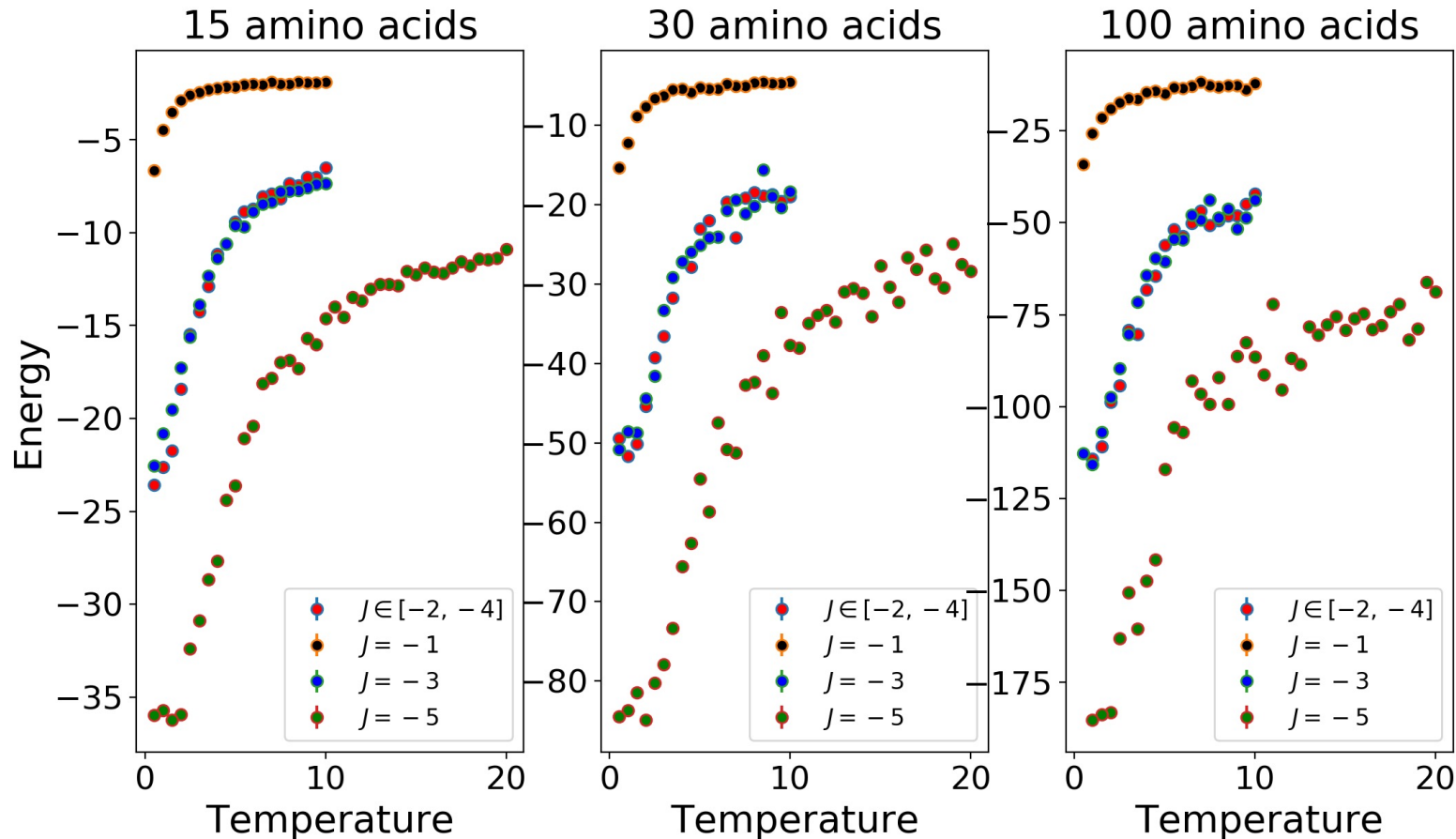


Figure 5: the relationship between temperature and the energy (left) and end-to-end distance (right) of the protein consisting of 15 amino acids with different interaction coefficient J .

- The role of J is to rescale the temperature: $e^{-\Delta E/k_B T} = e^{-J\Delta n/T}$
- Temperature:**
- In low T , the extent of folding tends to be high.
- There is a **phase transition** as raising temperature.
- When $T \uparrow$, the energy and length stop to increase and approach to **the certain value**.

Simulation Results: Effect of N

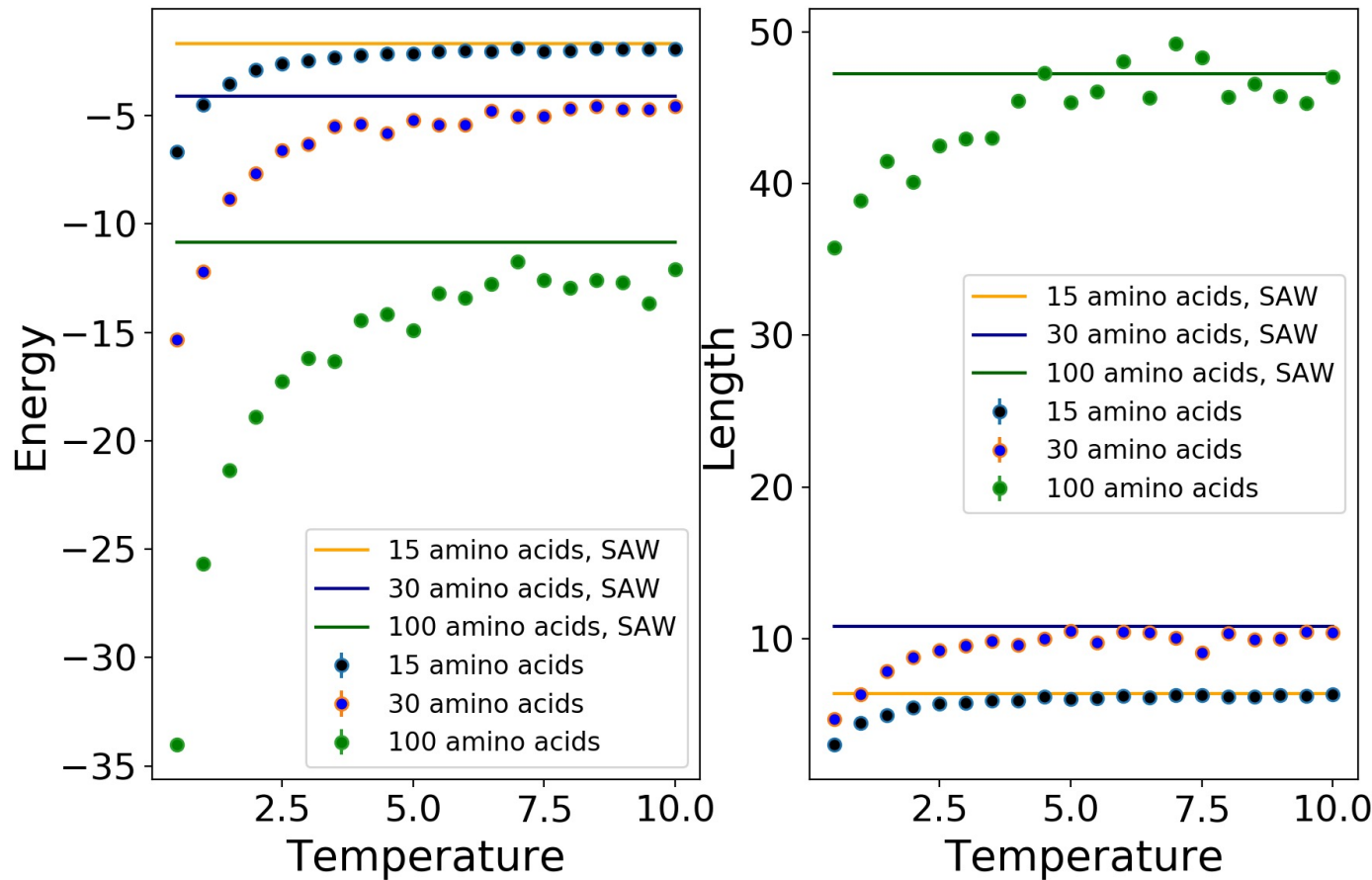


- Rules discussed before are **general**
- N **doesn't change** the transition temperature.

Figure 6: The relationship between temperature and the energy of the protein consisting of 15, 30, 100 amino acids with different interaction coefficient J .

The Certain Value: Self-Avoiding Random Walk

$$J = -1$$



- This rule is general and applies to all chains with different number of amino acids.
- The effect of thermal fluctuation dominates: $e^{-\Delta E/k_B T} \rightarrow 1$ (large T)

Figure 7: The comparison of SAW simulation and protein folding simulation for chains with 15, 30, 100 amino acids. The interaction coefficient J is -1.

Phase Transition

15 amino acids

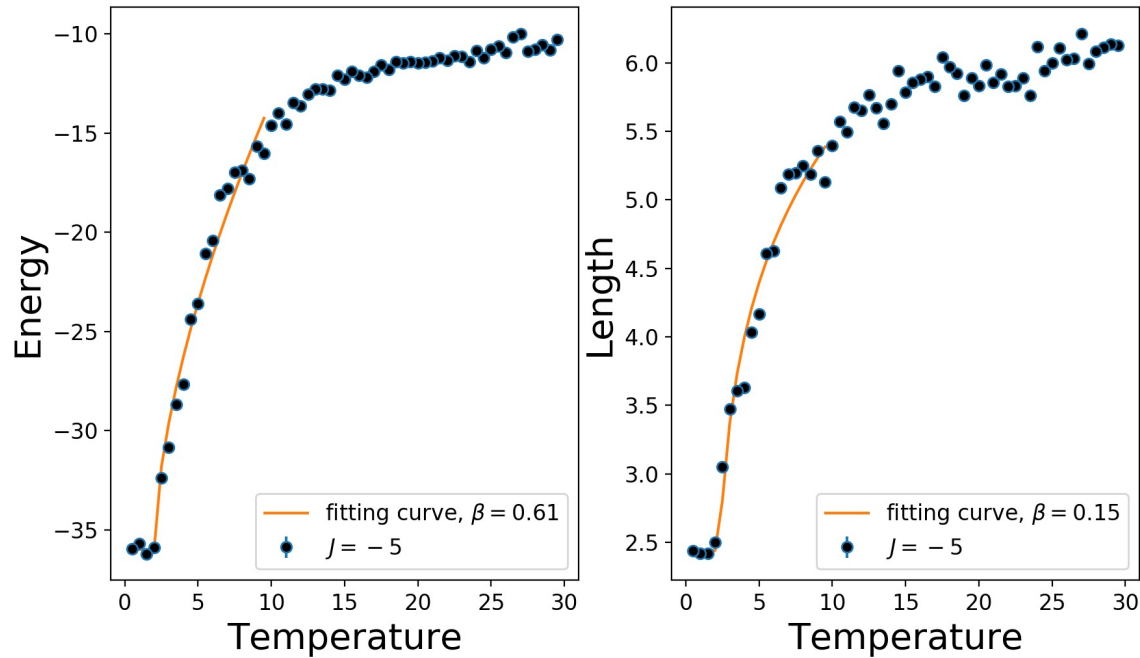


Figure 8: The fitting results of critical exponents for chains with 15 amino acids. The interaction coefficient J is -5.

Fitting function: $y = a\left(\frac{T}{T_c} - 1\right)^\beta$

- **Critical exponent:** $k = \lim_{\tau \rightarrow 0} \frac{\log(|f(\tau)|)}{\log(|\tau|)}$ [6]
- Reduced temperature: $\tau = \frac{T - T_c}{T_c}$
- This results in the power law: $f(\tau) \propto \tau^k$

15 amino acids

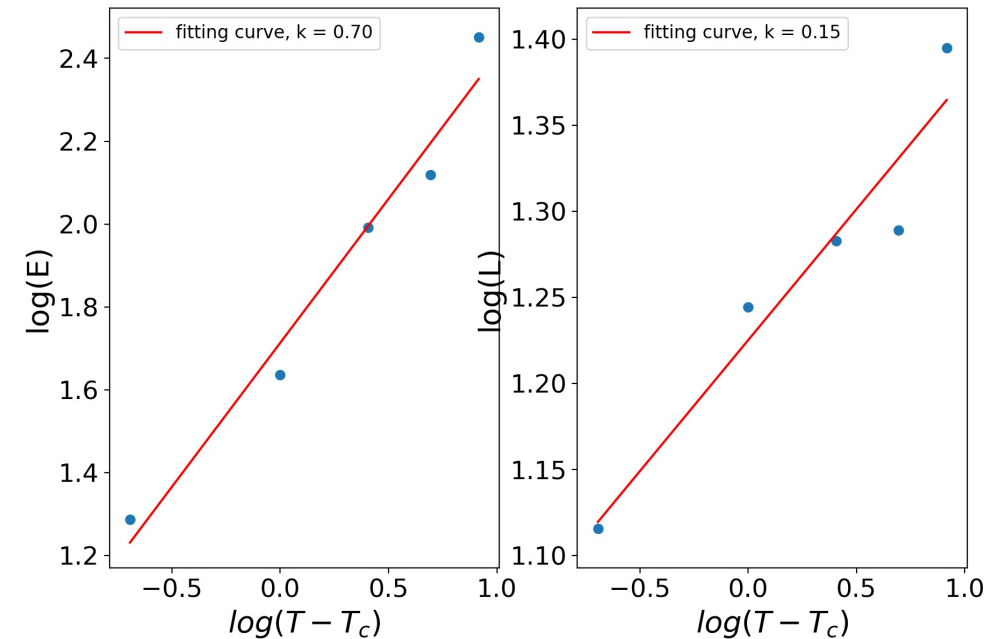


Figure 9: The linear fitting results of $\ln(E)$, $\ln(L)$ vs $\ln(T - T_c)$. (χ^2 plot shows that $T_c = 2$.)

Conclusion & Discussion

- 1) The extent of folding tends to be high in low temperature.
- 2) There is a phase transition as gradually changing temperature in the protein system. The transition temperature is independent of N .
- 3) The energy and the end-to-end distance approaches to the results of self-avoiding random walk when temperature is high enough, which indicates that the effect of thermal fluctuation dominates in high temperature.

Improvements:

- a) Increase the Monte Carlo step for $N=100$ case to get better curve.
- b) Further study is needed to explain the phase transition phenomena.

References:

- [1] Christian B. Anfinsen. The formation and stabilization of protein structure. *Biochemical Journal.*, 128(4):737–749, 1972.
- [2] Zejun Ding. Lecture Notes for Computational Physics. *manuscript*.
- [3] WK Hastings. Monte-Carlo Sampling Methods Using Markov Chains and Their Applications. *Biometrika*, 57(1):97–&, 1970.
- [4] S. Kirkpatrick, C. D. Gelatt, and M. P. Vecchi. Optimization by simulated annealing. *Science*, 220(4598):671–680, 1983.
- [5] Nicholas Metropolis, Arianna W. Rosenbluth, Marshall N. Rosenbluth, Augusta H. Teller, and Edward Teller. Equation of state calculations by fast computing machines. *The Journal of Chemical Physics*, 21(6):1087–1092, 1953.
- [6] https://en.wikipedia.org/wiki/Critical_exponent

Thank You!