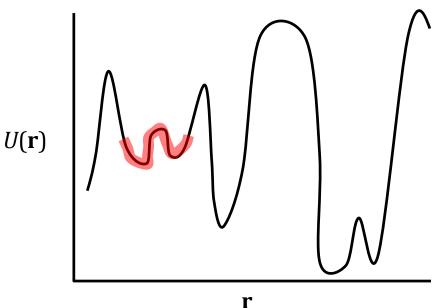
Modern Monte Carlo

Problems

- we are interested in properties at room temperature
- at room temperature, processes are slow
 - phase transitions, protein structure re-arrangement..
 - system can be trapped



• most large moves are rejected (wasted cpu time)

Goals

Speed simulation

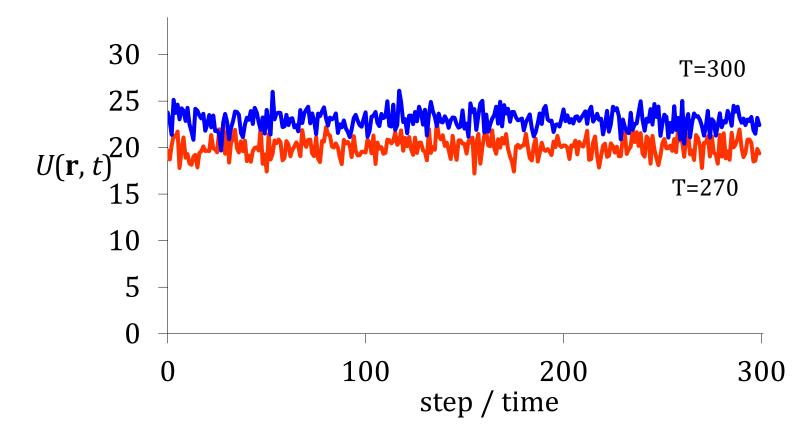
- two approaches
 - make barriers easier to pass
 - waste less time on failed moves

Restrictions

- must retain Boltzmann distribution
- must preserved detailed balance

Parallel Tempering / Replica exchange

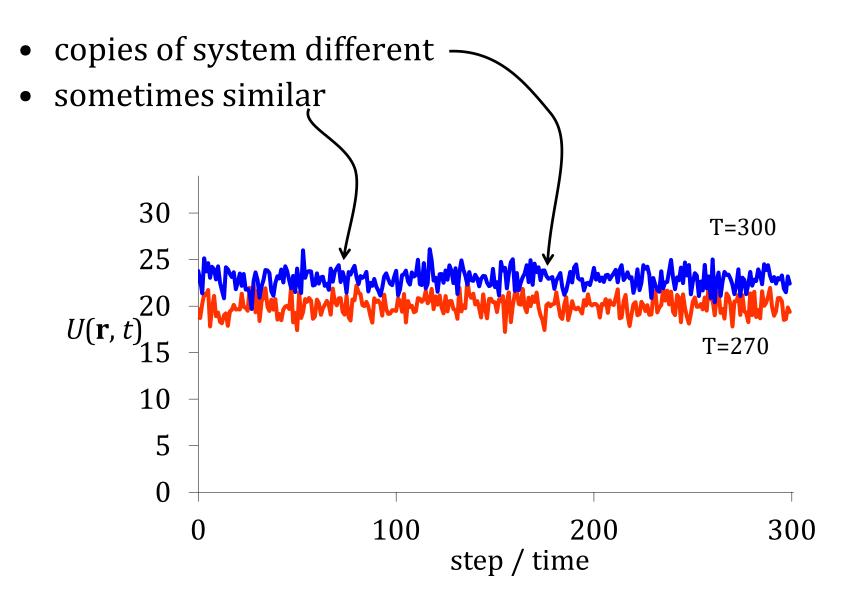
Two simulations, two temperatures



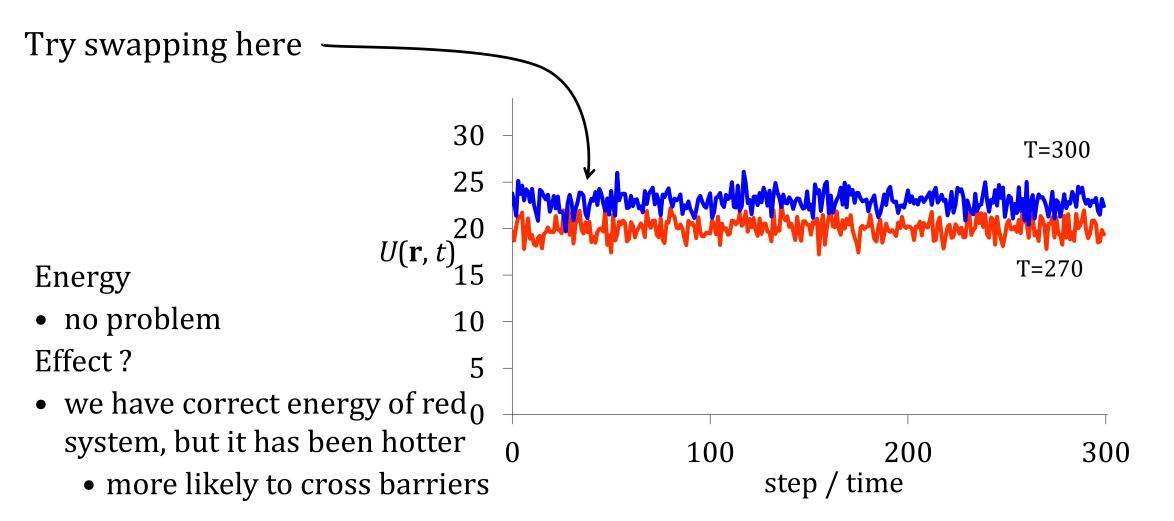
Hotter simulation moves faster, hops over barriers but

• it does not give $\langle A \rangle$ for desired temperature (270 K)

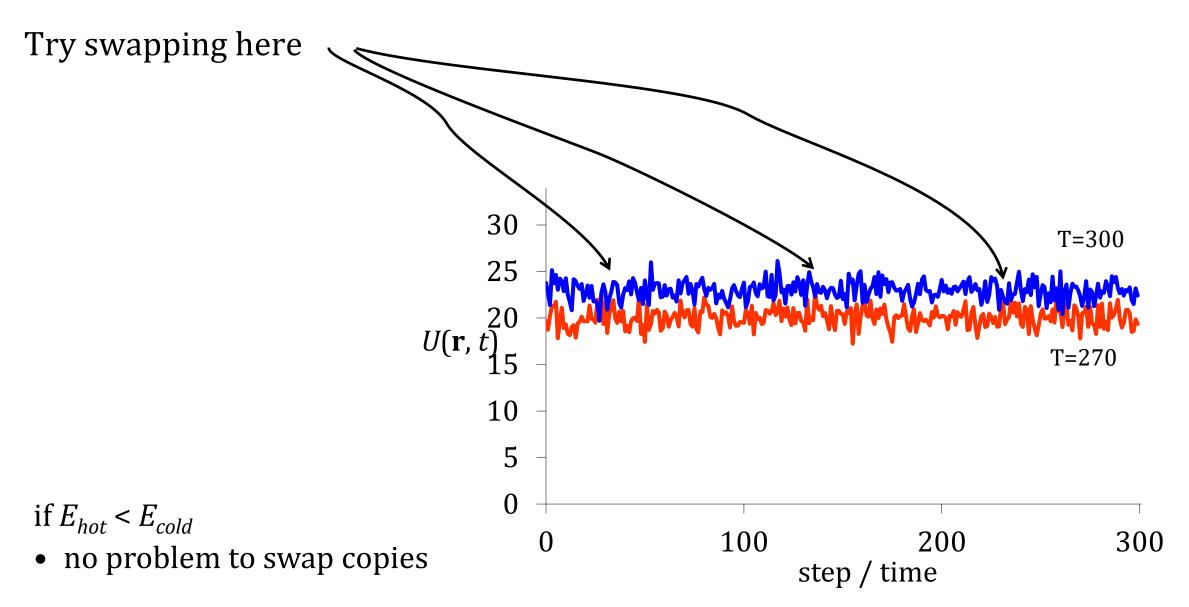
Closer temperatures



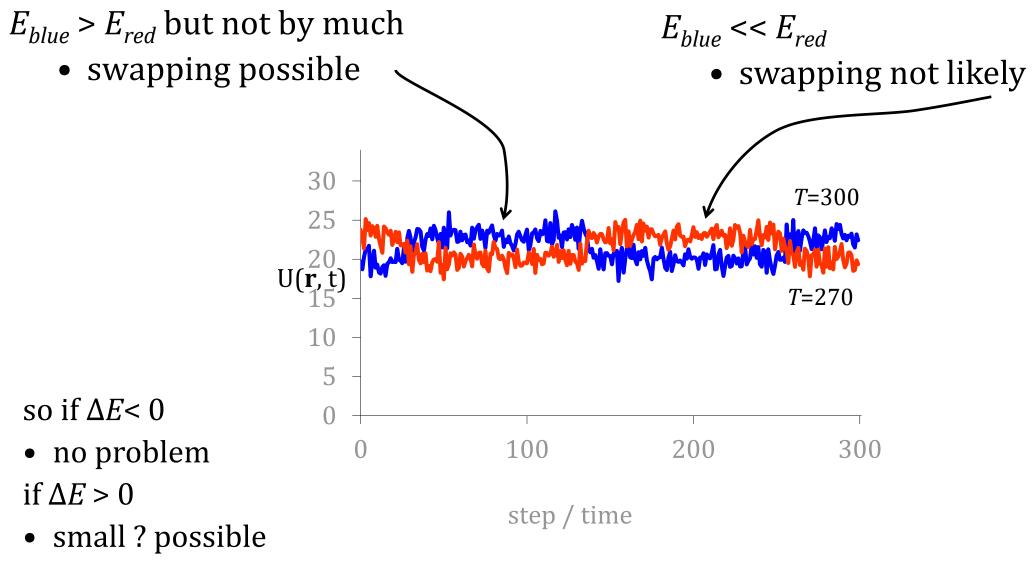
swaps of copies



easy swaps



possible swaps



• big ? less likely

Probability of a total system

• probability of one system
$$i \quad p_i = \frac{e^{-E_i/kT_i}}{Z_i}$$

• probability of whole system

$$p_{old} = p_i p_j = \frac{e^{-E_i/_{kT_i}}}{Z_i} \frac{e^{-E_j/_{kT_j}}}{Z_j}$$
$$= \frac{e^{\left(\frac{-E_i}{kT_i} + \frac{-E_j}{kT_j}\right)}}{Z_i Z_j}$$

- probability of system before and after swap ? $\frac{p_{new}}{p_{old}}$
- *Z*'s cancel

Exchange Probability

Question

- could the blue be part of the red ensemble?
- could the red be part of the blue ensemble? Depends on temperatures, ΔE

$$p_{swap} = \exp\left(\frac{E_j - E_i}{k(T_i - T_j)}\right)$$

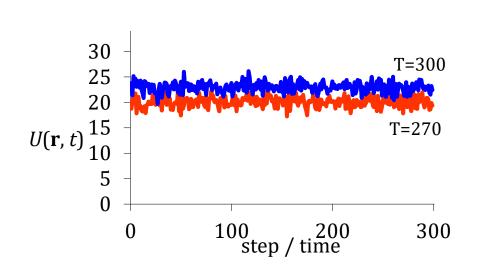
if $p_{swap} > 1$

if
$$p_{swap} >$$

accept

else use random number [0..1] and compare with p_{swap}

- consider $E_i \approx E_i$
- blue bit higher than red (moves likely) \bullet
- blue much higher than red (moves very unlikely) lacksquare



Implementing

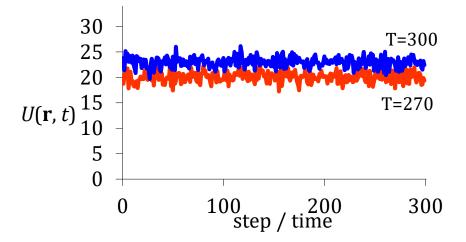
Example

- try 100 moves normal MC of each system
- try 1 exchange / swap of systems
- swap means:
 - in MC steps ($e^{-\Delta E/kT}$) change T_1 and T_2

Result

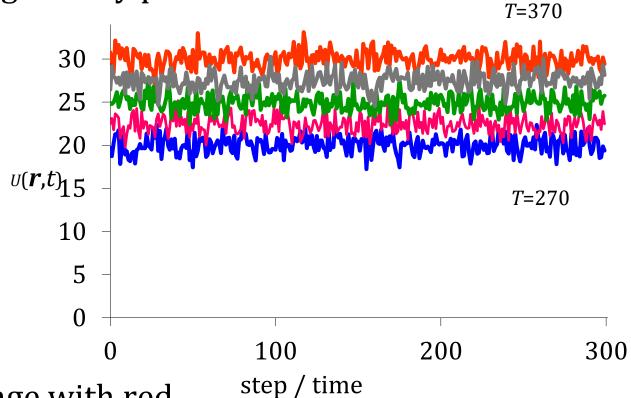
- two simulations
- each has Boltzmann distribution at right temperature
- cooler system has visited high temperatures / moved faster
- generalising

• ..



Many replicas

- run many copies, similar temperatures
- every *N* moves, attempt an exchange of any pair



- normally blue would never exchange with red
- now possible in several steps
- red simulation is a valid ensemble at T_{red}

Implementation

Any set of exchange attempts OK

• may not be efficient

Detail balance preserved

Easy to implement

- set up *N* simulations at different temperatures
- whenever a swap is successful, set T_i to T_j and T_j to T_i

Alternative perspective

- like simulated annealing but
 - annealing schedule (cooling) is automatic

Configurational Bias Monte Carlo Rosenbluth sampling

Many Monte Carlo methods

- do not take random step
- find a low energy direction
- trial move more likely in that direction
- make acceptance probability less likely

Result

- less time spent generating unlikely moves + energy calculation
 Rule
- must maintain detailed balance
- must finish with a Boltzmann distribution
- Example discrete system

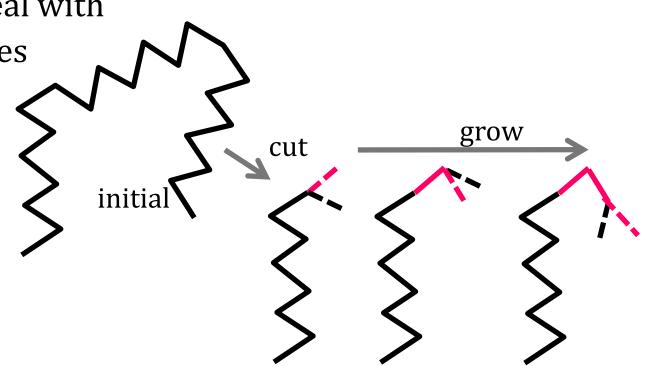
Discrete Models / Chain growth moves



- particles only exist in certain places
- can only occupy certain states Off-lattice discrete protein

Typical moves set

- pick random site in chain
- discard one half
- re-grow each site
- look at new configuration, accept/reject
- big reorganisation possible



Chain regrowth methods

Moves are big, but

• in a dense system, most will be rejected

We have big moves, but consider each step



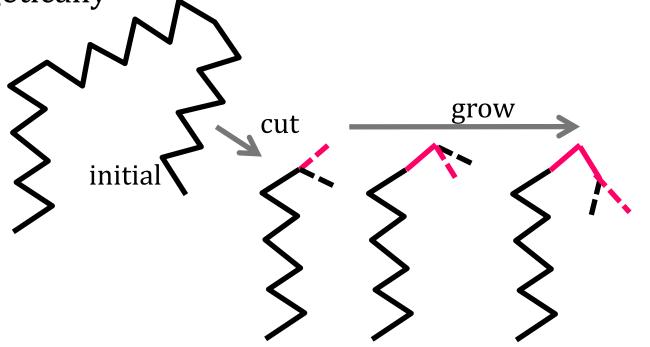
Looking at sub-moves

at first step

• one possible direction is more likely

what if we move in the more likely region?

- we will tend to move downhill energetically
 - no Boltzmann distribution
- move $N_i \pi(i \to j) \neq N_j \pi(j \to i)$
 - detailed balance not preserved



Bias

Make downhill moves more likely

- make them more difficult to accept Sometimes try uphill moves
- gain
 - fewer attempts at uphill moves
 - keep detailed balance + Boltzmann distribution

Next step

- do several biased moves
 - set of (probably) downhill moves

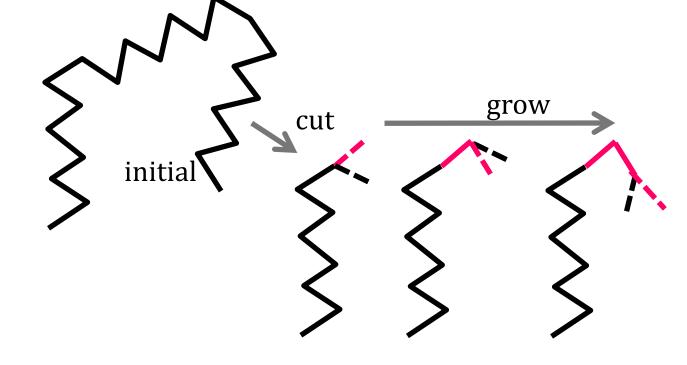
One step

Look at red and black choices

• calculate E_{black} , E_{red} and probabilities

$$p_{black} = \frac{e^{\frac{-E_{black}}{kT}}}{\sum_{i}^{red, black, \dots} e^{\frac{-E_{i}}{kT}}}$$

$$\sum_{i=1}^{N_{choices}} p_i = 1$$



- pick a direction according to *p_i*
- example...

direction picking

We have three possible directions

• $p_1 = 0.2, p_2 = 0.5, p_3 = 0.3$ from Boltzmann weights

| pick random number | $0 \leq x \leq 1$ |
|--------------------|-------------------|
| if $0 \le x < 0.2$ | choose (1) |
| elseif $x < 0.7$ | choose (2) |
| else | choose (3) |

- what have we got now ? not much yet
- usually choose single steps and preserve Boltzmann distribution

formalism

Where we have N_{choice} possible directions

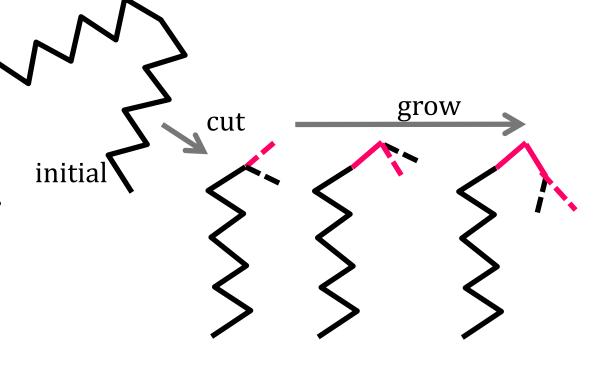
$$p_i = \frac{e^{\frac{-E_i}{kT}}}{w}$$

$$w = \sum_{j=1}^{N_{choice}} e^{\frac{-E_j}{kT}}$$

w will come back in a moment

Several Bias steps

- break chain
- pick first step with bias
- second step with bias
- ...
- chain complete
- Heavily biased
- series of N_{step} steps usually favourable
 - without accept / reject along the way
- how to correct?
 - introduce "Rosenbluth factor"
 - W_o (old), W_n (new / trial)

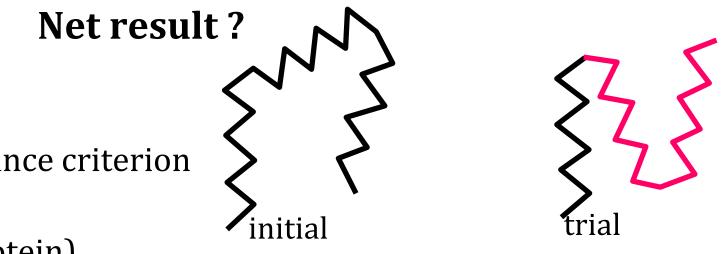


Rosenbluth factor



Rosenbluth factor W_o

- pretend that the chain was chopped and calculate w_m for each step Accept reject
- if $W_n/W_o > 1$ accept
- else accept with $p = W_n / W_o$



- take *N*_{step} biased moves
- fix up distribution via acceptance criterion
- Practical explanation (dense protein)
- each step we put atoms in a likely place (not on top of other atoms)
- after N_{steps} we have a chain which is probably physically likely (unlikely to waste time on crazy moves)
- Compare with normal Monte Carlo
- to go from black to red would have required a very specific set of random moves (unlikely to be found)

Who uses configurational biased MC?

- proteins, polymers
- easiest when system is discrete
 - difficult to code in continuous systems
- typical of many methods (introduce bias and correct afterwards)
- putting techniques together

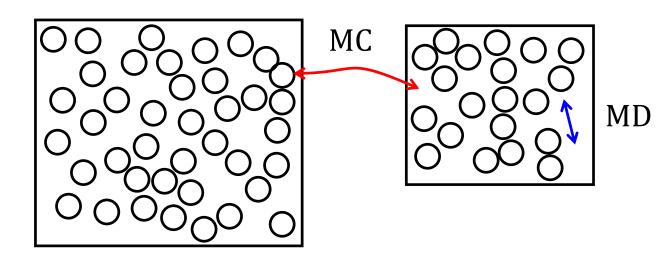
Combinations of techniques

Goal

- finish with a Boltzmann distribution
- dynamics ? maybe

Combinations

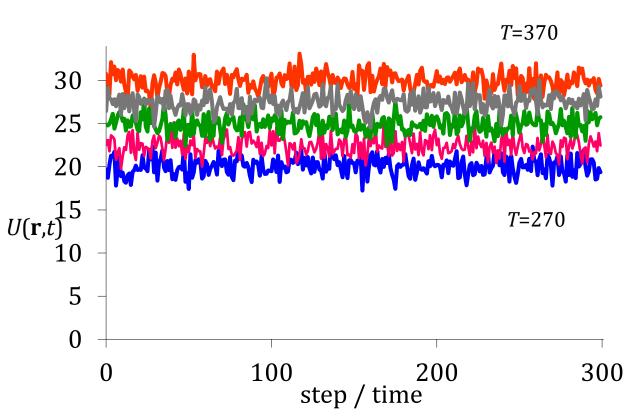
- Molecular dynamics and Monte Carlo ?
- Monte Carlo good for non-physical systems



More combinations

Replica Exchange method

- MC or MD
- both will give ensemble / distribution at desired temperature



Imagine

- MC is good for complete re-arrangement of chain
- MD explores local (nearby) configurations
- could combine biased MC with MD

Comparison with other methods

- classic minimisation method genetic algorithm
- basic idea
 - 100 or 1000 copies of system (protein, travelling salesman routes)
- make 100 copies of system

 while (not happy)
 find 50 worst copies (highest energy) throw away
 copy 50 best
 for (*i* = 0; *i* < 50; *i*++)
 apply random changes, combine copies
- system will gradually improve fittest copies are kept

Comparing to MC

- Methods like genetic algorithm work with unknown distribution
- no theory to fall back on
 - no defined temperature
 - no defined probabilities

Summary of everything

Methods like molecular dynamics /Monte Carlo

- infinite number of variations possible / legal
- best may be system dependent
- not restricted to molecular / atomic systems
- Arbitrary decisions
- temperature, move types

Questions

How happy are we with replica exchange?

- I simulate with four replicas $4 \times$ the time
- Could I just simulate 4 times longer ?

We do Monte Carlo or dynamics simulation on a system

- explain the difference in philosophy when we want to
 - simulate the world
 - optimise the system

More questions

- Think of
- 1. score function
- 2. move set for
- Monte Carlo for sequence optimisation ? (make a better protein sequence)
- Monte Carlo for sequence alignments ?
 - why would you want to ?
 - what might be the problems ? (describe the energy landscape)
- why genetic algorithms are evil