

16a) Example of Monte-Carlo algorithm

Metropolis algorithm

Spin configuration $c_{n,i}$ $i=1, \dots, N$
 $c_{n,i} = S_i$

i) do local update at site i

$$c'_i = c_{n,i} + \delta c_i$$

$$c'_j = c_{n,j} \text{ for } j \neq i$$

eg. For the Ising model δc_i is
 a spin flip at site i

ii) Choose a random $x \in [0, 1]$

then

$$c_{n+1,i} = c_{n,i} \text{ if } e^{-\beta H(c') + \beta H(c)} < x$$

$$c_{n+1,i} = c'_i \text{ if } e^{-\beta H(c') + \beta H(c)} > x$$

If done for all lattice sites start again
 at the beginning

16 Monte Carlo simulations

In many fields of physics we have to evaluate multidimensional integrals

- Examples:
- Lattice QCD
 - Statistical Mechanics
 - Green's function Monte-Carlo in solid state and nuclear physics

Riemann integration: # points is N^d for a d -dimensional integral

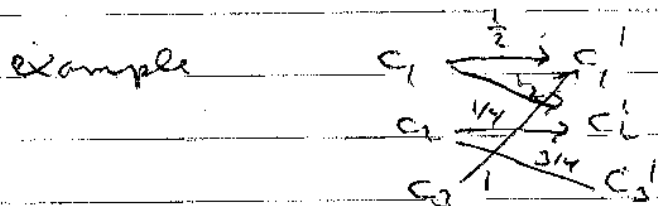
Monte Carlo simulation: Choose N points at random, then the error is $\propto 1/\sqrt{N}$

problem: Typically the integrand is negligible in most of its domain. We wish to choose the points mainly in the region where the integrand is large. This is the aim of Monte-Carlo algorithms

~ Theorem: $\sum_{c'} T(c, c') = 1$

a state has to go somewhere

but $\sum_c T(c, c') \neq 1$ in general



We define $P_n(c)$ as the probability to find the configuration c after n steps

Evolution $P_{n+1}(c) = \sum_c T(c, c') P_n(c)$

stationary solution: $\bar{p}(c) = \sum_{c'} T(c, c') \bar{p}(c')$

Theorem: In a Markov process with detailed balance $\bar{p}(c) = e^{-\beta H(c)}$ is a stationary solution

~ Proof: sum over c' in detailed balance

$$e^{-\beta H(c)} = \sum_{c'} T(c, c') e^{-\beta H(c')} \quad \left(\sum_{c'} T(c', c) = 1 \right)$$

~ This theorem is the basis of many Monte-Carlo algorithms. We have to construct a Markov chain that satisfies detailed balance.

Sometimes it is useful to have an evolution by a Hermitian matrix

Define $u_n(c) = p_n(c) e^{\frac{\beta}{2} H(c)}$

then $u_{n+1}(c') = \sum_c \underbrace{e^{-\frac{\beta}{2} H(c)} T(c, c')}_{R(c, c')} e^{\frac{\beta}{2} H(c)} u_n(c)$

~ $R(c, c') = e^{-\frac{\beta}{2} H(c)} T(c, c') e^{\frac{\beta}{2} H(c')}$

$= e^{-\frac{\beta}{2} H(c)} \frac{e^{-\beta H(c)}}{e^{-\beta H(c')}} e^{\frac{\beta}{2} H(c')}$

$= R(c', c)$

R is Hermitian \Rightarrow For $n \rightarrow \infty$ u_n converges to the eigenvector with the largest eigenvalue (if it is nondegenerate)

~ $u_{eq}(c) = e^{-\beta H(c)/2}$ is an eigenvector of $R(c, c')$, corresponding to $p_{eq}(c) = e^{-\beta H(c)}$