Model Markov Chain Monte Carlo Possible Errors Further goal

Lattice Geometry Lattice Sites Magnetization Energy

Sampling the Ising Model

Abraham Hinteregger

University of Vienna

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History

- Proposed by Wilhelm Lenz to his student Ernst Ising
- ▶ 1924: Ernst Ising Beitrag zur Theorie des Ferromagnetismus¹
 "Es entsteht ...[durch] ... die Beschränkung der
 Wechselwirkung auf diejenige benachbarter Elemente
 [...] kein Ferromagnetismus."
- ▶ 1936: Rudolph Peierls On Ising's model of ferromagnetism²

 "[...] for sufficiently low temperatures the Ising

 model in two [or more] dimensions shows

 ferromagnetism [...].

¹Zeitschrift für Physik Februar-April 1925, Volume 31, Issue 1, pp 253-258

²Cambridge Philosophical Society 1936, Volume 32, Issue 03, Oct.

Lattice Geometry Lattice Sites Magnetization Energy

Lattice

Figure: Square Lattice in 1 dimension

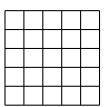


Figure: Square Lattice in 2 dimensions

Lattice Geometry Lattice Sites Magnetization Energy

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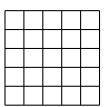


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Lattice Geometry Lattice Sites Magnetization Energy



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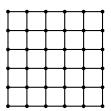


Figure: Square Lattice in 2 dimensions



Figure: Square Lattice in 1 dimension

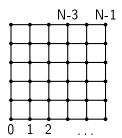


Figure: Square Lattice in 2 dimensions



Figure: Square Lattice in 1 dimension

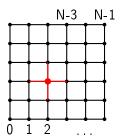


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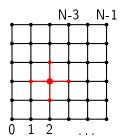


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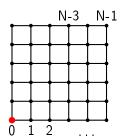


Figure: Square Lattice in 2 dimensions



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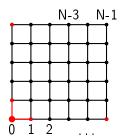
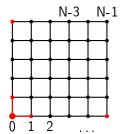


Figure: Square Lattice in 2 dimensions



Figure: Square Lattice in 1 dimension



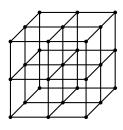
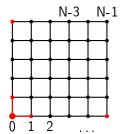


Figure: Square Lattice in 2 and 3 dimensions



Figure: Square Lattice in 1 dimension



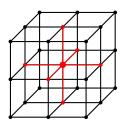


Figure: Square Lattice in 2 and 3 dimensions

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Lattice Sites

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Lattice Sites

- Each site has a state $s_i = -1 \lor 1$
- Assignment of states $S = (s_0, s_1, s_2, \dots, s_{N-1})$ to the lattice sites is called a configuration
- ► Therefore 2^N unique configurations for a lattice with N lattice sites.

Magnetization

► The magnetization of a configuration is calculated by

$$m_{x}=m(S_{x})=\sum_{i}^{N}s_{i}\in[-N,N]$$

usually the magnetization per spin is considered

$$M_{\mathsf{x}} = \frac{m_{\mathsf{x}}}{\mathsf{N}} \in [-1, 1]$$

Energy

► Each configuration has a corresponding energy - the Hamiltonian.

$$E_x = E(S_x) = H(s_0, s_1, \dots, s_{N-1}) = -J \sum_{\langle i,j \rangle} s_i \cdot s_j - h \sum_i s_i$$

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Figure: Energy contribution (nearest neighbor interaction) of the bonds connected to the central lattice site

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Boltzmann distribution

Probability of system being in the state S is given by the Boltzmann distribution ($\beta = 1/kT$)

$$p_X = p(S_X) = \frac{e^{-\beta E_X}}{Z}$$

Z is the the partition function

$$Z = \sum_{i}^{2^{N}} e^{-\beta E_{i}}$$

Model Markov Chain Monte Carlo Possible Errors Further goal

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- ► Sampling from random configurations would lead to many high- energy (temperature) configurations (simple sampling)
- \rightarrow importance sampling

Markov Chain

- ightharpoonup Chain of iteratively created configurations $C_1, C_2, \ldots C_n$
- ► Resulting configurations correspond to the desired probability distribution *p* and span the entire state space.
- ▶ Configuration C_t only depends on C_{t-1} (Markov property).

Transitions in the Markov Chain

Probability for being in state A: p_A Transition probability for transition $S_A \rightarrow S_B$: p_{AB}

If it fulfills detailed balance (it must!)

$$p_A \cdot p_{AB} = p_B \cdot p_{BA}$$

the following relation for the transition probability follows:

$$\frac{p_{AB}}{p_{BA}} = \frac{p_B}{p_A} = \left(\frac{Z}{Z}\right) \frac{e^{-\beta E_B}}{e^{-\beta E_A}} = e^{-\beta (E_B - E_A) = e^{-\beta \Delta E}}$$

Metropolis- Hastings Algorithm

- ightharpoonup Configuration after t steps is C_t
- ▶ Flip one lattice site $\rightarrow C'_t$
 - ▶ has to be chosen randomly suitable RNG necessary
- Calculate energy difference $\Delta E = E'_t E_t$
- Calculate acceptance probability P

$$P = \min \left(1, e^{-eta \cdot \Delta E}
ight), \qquad eta = 1/kT > 0$$

- Generate random number $r \in [0,1[$
 - $ightharpoonup r < P \rightarrow C_{t+1} = C'_t$
 - $ightharpoonup r > P
 ightharpoonup C_{t+1} = C_t$

Magnetization over time (MH- Algorithm)

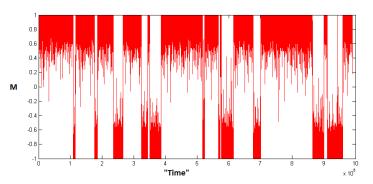


Figure: Not in physical time but Monte Carlo time steps.

Swendsen- Wang Algorithm

- ► Search for clusters with equal spins
- Generate a bond between lattice sites in such a cluster with

$$p = 1 - e^{-2/T}$$

- ► Search for clusters connected by bonds
- Assign a random spin to each cluster.

Magnetization over time (SW- Algorithm)

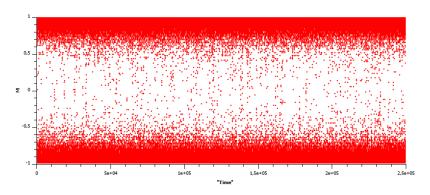
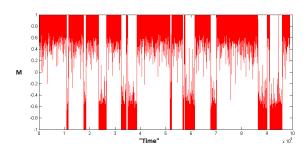


Figure: Not in physical time but Monte Carlo time steps.

Autocorrelation

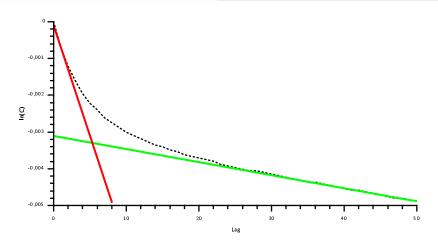
▶ Similarity of a function with a "lagged" version of itself.

$$C_M(t) = \frac{cov(M_n, M_{n+t})}{std(M_n)std(M_{n+t})} = \sum_i c_i e^{-t/\tau_i}$$



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Autocorrelation Effective chain length Finite Size Effects Overview



$$ln\left(C_{M}(t)
ight) = \sum_{i} \left(ln(c_{i}) - rac{t}{ au_{i}}
ight)$$

Effective Markov- chain length

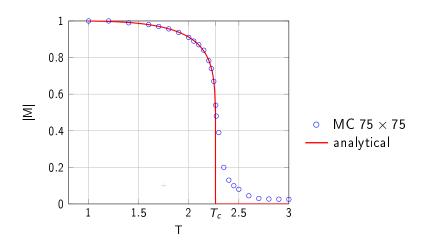
naive error analysis leads to underestimated errors due to autocorrelation.

$$var(M) = \sigma_M^2 = \frac{\sum_i (M_i - \bar{M})^2}{n}$$

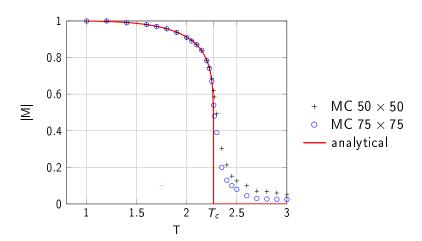
realistic error considers autocorrelation with smaller sample-size n_{eff}:

$$n_{eff} = \frac{n}{2\tau_{int}} \approx \frac{n}{2\sum_{i} c_{i}\tau_{i}}$$

Absolute magnetization per spin



Absolute magnetization per spin



Autocorrelation Effective chain length Finite Size Effects Overview

- ▶ Update algorithm must be chosen according to parameters.
- ► Sample size must be chosen big enough to compensate autocorrelation
- Right size of the lattice to adjust for finite size effects depending on temperature.

Ambition Kawasaki Dynamics Current Potts Model

Ambition

- ► Simulating flow of sediments in current
- ► Needed adjustments:
 - constant "spin" ratio and...
 - flowing and not random spawning/despawning of "particles"
 - current in form of outer force
 - more states for a lattice site

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- ► Simulating flow of sediments in current
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 - ightharpoonup constant "spin"- ratio and... ightharpoonup Kawasaki Dynamics
 - flowing and not random spawning/despawning of "particles"
 - ightharpoonup current in form of outer force ightharpoonup force term in Hamiltonian
 - ightharpoonup more states for a lattice site ightharpoonup Potts Model

Kawasaki Dynamics

- Choose A-B bond
- ▶ Calculate ΔE for $A B \rightarrow B A$
- Accept with probability depending on energy difference

Behaviour:

- ► Spin- ratio (Magnetization) constant.
- ▶ Different spins get sorted (if coupling constant J is positive).

Current

$$H_G = -J \sum_{\langle i,j \rangle} s_i s_j - \sum_j h_j \sum_{\mathit{line}_j} s_i$$

with

$$h_j = h_1 + j \frac{(h_L - h_1)}{L}$$

Current

$$H_G = -J \sum_{\langle i,j
angle} s_i s_j - \sum_j h_j \sum_{\mathit{line}_j} s_i$$

with

$$h_j = h_1 + j \frac{(h_L - h_1)}{L}$$

Potts Model

- Generalized version of the Ising model.
- ▶ states not only $-1 \land 1$ but (discrete) angles.
- ▶ Energy usually in the form $H = -J_c \sum_{i,j} cos(\theta_i \theta_j) + \dots$

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Thanks for your attention

- ► Sourcefiles and binaries on my github https://github.com/oerpli/Ising
 - currently only tested on Windows 7& 8.
 - Swendsen Wang Algorithm implemented with recursive DFS accordingly stackoverflow error with big clusters (depending on configuration)