

Critical behaviour of the surface tension in the 3D Ising model

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Summary

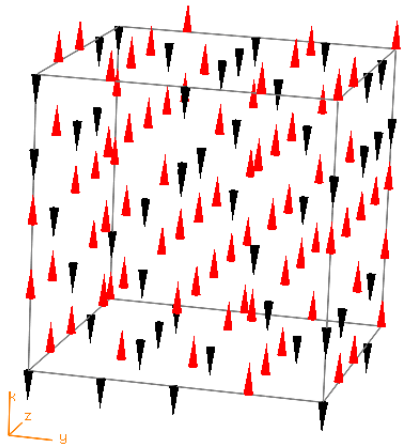
- 3D Ising models
- Definition of the surface tension
- Cluster algorithms and boundary flip
- (Notes on the implementation?)
- Estimation of the errors and autocorrelation
- Fit of the free energy
- Fit of the critical behaviour
- (Conclusion?)

3D Ising model

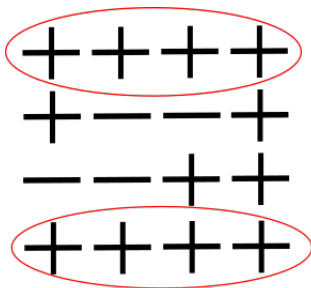
$$\mathcal{H} = - \sum_{\langle x,y \rangle} J_{\langle x,y \rangle} s_x s_y$$

$J_{\langle x,y \rangle} = 1$ ferromagnetic

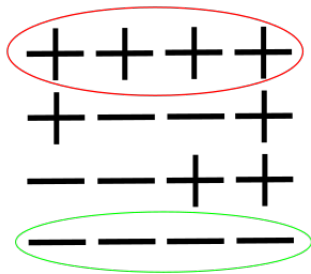
$J_{\langle x,y \rangle} = -1$ antiferromagnetic



Definition of the surface tension

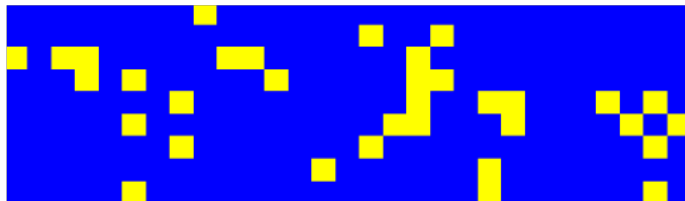


Z_{++}



Z_{+-}

$$\sigma = -\lim_{L \rightarrow \infty} \frac{1}{L^2} \log \frac{Z_{+-}}{Z_{++}} \quad L \times L \times T, \quad T = cL$$


 Z_{+-}

 Z_{++}

$$\sigma = - \lim_{L \rightarrow \infty} \frac{1}{L^2} \log \frac{Z_{+-}}{Z_{++}} = \lim_{L \rightarrow \infty} \frac{1}{L^2} (F_{+-} - F_{++}) = \lim_{L \rightarrow \infty} \frac{F_s}{L^2}$$

σ = interface free energy per unit area

Redefinition of Z_{++} and Z_{+-} .

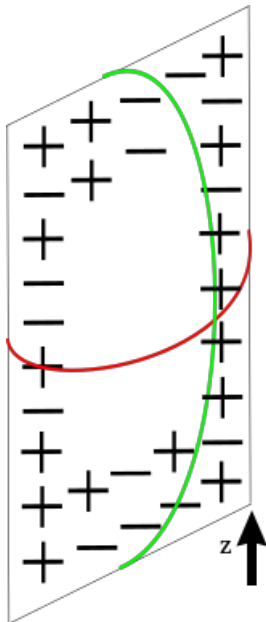
$Z_{++} \rightarrow$ ferromagnetic link
between **top and bottom**.

$Z_{+-} \rightarrow$ **antiferromagnetic** link
between **top and bottom**.

Always ferromagnetic link in x
and y directions.

Same definition for σ .

Periodic boundary conditions
reduce the finite size effect.



Montecarlo simulations can't measure Z !

Solution: $J_{\langle x,y \rangle}$ between top and bottom becomes a **dinamical variable** that is summed over in Z .

$J_{\langle x,y \rangle} = 1$ (periodic b.c.) $J_{\langle x,y \rangle} = -1$ (antiperiodic b.c.)
Other $J_{\langle x,y \rangle}$ remains ferromagnetic.

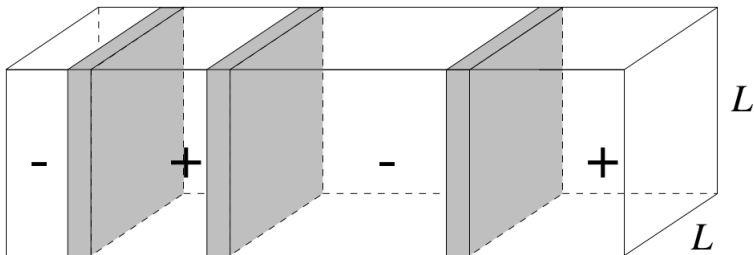
$$Z = \sum_{\{s\}, J} \exp \left(\beta \sum_{\langle x,y \rangle} J_{\langle x,y \rangle} s_x s_y \right)$$

$$\frac{Z_{+-}}{Z_{++}} = \frac{\frac{Z_{+-}}{Z}}{\frac{Z_{++}}{Z}} = \frac{\langle \delta_{J=-1} \rangle}{\langle \delta_{J=+1} \rangle}$$

Ratio of measurable expectation values.

We redefine the free energy of the interface in order to improve the convergence properties of $\frac{F_s}{L^2}$ to σ when $L \rightarrow \infty$.

Thermodynamic limit \rightarrow only **one** interface
For finite L multiple interface can be present. An even number for Z_{++} and odd for Z_{+-} .



F_s is the free energy of a single surface. There are $\sim T$ different position for the interface.

$$Z_1 = T \exp(-F_s) = \exp(-F_s + \ln T)$$

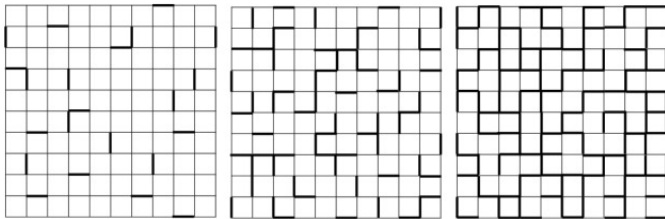
$$\frac{Z_{+-}}{Z_{++}} = \frac{Z_1 + \frac{Z_1^3}{3!} + \frac{Z_1^5}{5!} + \dots}{1 + \frac{Z_1^2}{2!} + \frac{Z_1^4}{4!} + \dots} = \tanh(\exp(-F_s + \ln T))$$

$$F_s = \ln(T) - \ln\left(\frac{1}{2} \ln\left(\frac{1 + \frac{Z_{+-}}{Z_{++}}}{1 - \frac{Z_{+-}}{Z_{++}}}\right)\right) \quad \sigma = \lim_{L \rightarrow \infty} \frac{F_s}{L^2}$$

Cluster algorithms and boundary flip

Cluster algorithms allow for simultaneous updates of large parts of the lattice. Thus reducing the autocorrelation time and the critical slowing down. Swendsen and Wang (1987).

Introduce link variables $\sigma_{\langle x,y \rangle} = \{0, 1\}$ on the lattice:



$$Z = \sum_{\{s=\pm 1\}} \exp \left(\beta \sum_{\langle x,y \rangle} s_x s_y \right) = \sum_{\{s=\pm 1\}} \prod_{\langle x,y \rangle} e^{\beta s_x s_y} =$$

$$= e^{-dV\beta} \sum_{\{s=\pm 1\}} \prod_{\langle x,y \rangle} (1 + \delta_{s_x, s_y} (e^{2\beta} - 1)) =$$

$$= e^{-dV\beta} \sum_{\{s\}} \prod_{\langle x,y \rangle} \sum_{\{\sigma_{\langle x,y \rangle}=0,1\}} [(1 - \sigma_{\langle x,y \rangle}) + \sigma_{\langle x,y \rangle} \delta_{s_x, s_y} (e^{2\beta} - 1)]$$

Also valid for generic coupling $J_{\langle x,y \rangle}$:

$$\begin{aligned}
 Z &= \sum_{\{s=\pm 1\}} \exp \left(\beta \sum_{\langle x,y \rangle} J_{\langle x,y \rangle} s_x s_y \right) = \sum_{\{s=\pm 1\}} \prod_{\langle x,y \rangle} e^{\beta J_{\langle x,y \rangle} s_x s_y} = \\
 &= e^{-dV\beta} \sum_{\{s=\pm 1\}} \prod_{\langle x,y \rangle} \left(1 + \delta_{J_{\langle x,y \rangle} s_x s_y, 1} (e^{2\beta} - 1) \right) = \\
 &= e^{-dV\beta} \sum_{\{s\}} \prod_{\langle x,y \rangle} \sum_{\{\sigma_{\langle x,y \rangle}=0,1\}} (1 - \sigma_{\langle x,y \rangle}) + \\
 &\quad + \sigma_{\langle x,y \rangle} \delta_{J_{\langle x,y \rangle} s_x s_y, 1} (e^{2\beta} - 1)
 \end{aligned}$$

For a fixed spin configuration $\{s\}$ the links are independent.

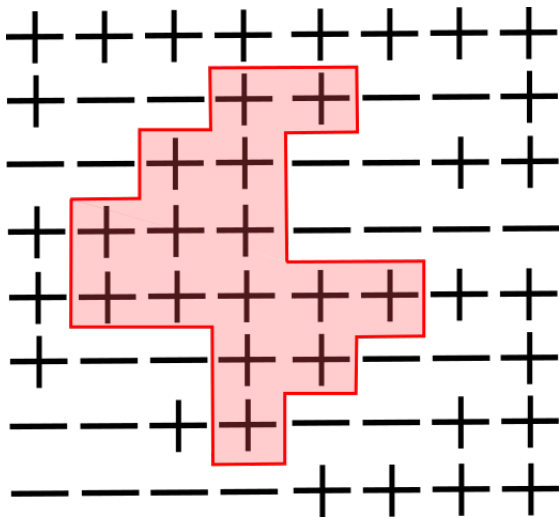
$$p_0 = p(\sigma_{\langle x,y \rangle} = 0) = \exp\left(-2\beta\delta_{J_{\langle x,y \rangle} s_x s_y, 1}\right)$$

$$p_1 = p(\sigma_{\langle x,y \rangle} = 1) = 1 - p(\sigma_{\langle x,y \rangle} = 0)$$

(if $J_{\langle x,y \rangle} s_x s_y = 1$ the weights in Z are normalized to $e^{2\beta}$)

For simplicity let's put $J_{\langle x,y \rangle} = 1$. For fixed $\sigma_{\langle x,y \rangle}$ only configurations of spins that satisfy the constraint $s_x = s_y$ where $\sigma_{\langle x,y \rangle} = 1$ have a non zero probability. All configurations of spin that satisfy the constraint have the same weight.

Definition: a **cluster** is a set of spins in the lattice path-connected by links with $\sigma_{\langle x,y \rangle} = 1$. If $J_{\langle x,y \rangle} = 1$ all sites of a cluster are forced to have the same spin.



Swendsen and Wang algorithm:

- Generate a link configuration $\sigma_{\langle x,y \rangle}$ based on the current spin configuration by using probabilities p_0 and p_1 .
- For each cluster choose a spin ($s = \pm 1$) with probability $\frac{1}{2}$. In general the update step must be compatible with the constraint $J_{\langle x,y \rangle} s_x s_y = 1$.
- The newly generated spin configuration is the next element of the Markov chain.

The SW is **ergodic**. We know prove that it satisfies the **detailed balance**.

In the cluster algorithm we update both the spins $\{s\}$ and the links $\{\sigma_{x,y}\}$. The ensemble contains both spin and link configurations: $\{s, \sigma\}$.

$$\frac{P(\{s_0, \sigma_0\} \rightarrow \{s_1, \sigma\})}{P(\{s_1, \sigma_0\} \rightarrow \{s_0, \sigma\})} = \frac{P(\{s_1\}|\{\sigma, s_0\}) P(\{\sigma\}|\{s_0\})}{P(\{s_0\}|\{\sigma, s_1\}) P(\{\sigma\}|\{s_1\})}$$

(this is not the detailed balance in the ensemble of $\{s, \sigma\}$ as the link configurations don't get exchanged!)

$P(\{s\}|\{\sigma, s_0\}) = \frac{1}{2^{\#cluster}}$ if $\{s\}$ is compatible with $\{\sigma\}$, null otherwise. Notice that in the SW algorithm $\{s\}$ is actually independent on $\{s_0\}$.

$$\text{Thus } P(\{s_1\}|\{\sigma, s_0\}) = P(\{s_0\}|\{\sigma, s_1\}).$$

In general this is the condition we ask to cluster algorithms.

$$P(\{\sigma\}|\{s\}) = \prod_{\substack{\sigma_{\langle x,y \rangle}=0 \\ J_{\langle x,y \rangle} s_x s_y=1}} e^{-2\beta} \prod_{\sigma_{\langle x,y \rangle}=1} (1 - e^{-2\beta})$$

for $\{s\}$ compatible with $\{\sigma\}$.

The first factor arise from the unconnected links for which $J_{\langle x,y \rangle} s_x s_y = 1$ each being in this state with probability $e^{-2\beta}$.

The second one is from the connected links (for which $J_{\langle x,y \rangle} s_x s_y = 1$ necessarily).

The second factor is **independent of $\{s\}$** and will be neglected in the sequent.

We obtain:

$$P(\{\sigma\}|\{s\}) = \prod_{\substack{\sigma_{\langle x,y \rangle}=0 \\ J_{\langle x,y \rangle} s_x s_y=1}} e^{-2\beta}$$

Now we compute (reminding $\{s_0\}$ and $\{s_1\}$ share the same $\{\sigma\}$):

$$\frac{e^{-\beta\mathcal{H}(\{s_1\})}}{e^{-\beta\mathcal{H}(\{s_0\})}} = \frac{\prod_{\langle x,y \rangle} e^{J_{\langle x,y \rangle} s_x^1 s_y^1}}{\prod_{\langle x,y \rangle} e^{J_{\langle x,y \rangle} s_x^0 s_y^0}} =$$

$$\frac{\prod_{J_{s_x^1 s_y^1}=+1}^{\sigma=0} e^{\beta} \prod_{J_{s_x^1 s_y^1}=-1}^{\sigma=0} e^{-\beta} \prod_{J_{s_x^1 s_y^1}=+1}^{\sigma=1} e^{\beta}}{\prod_{J_{s_x^0 s_y^0}=+1}^{\sigma=0} e^{\beta} \prod_{J_{s_x^0 s_y^0}=-1}^{\sigma=0} e^{-\beta} \prod_{J_{s_x^0 s_y^0}=+1}^{\sigma=1} e^{\beta}}$$

The last part having $\sigma = 1$ obviously depends only on $\{\sigma\}$ for all $J_{\langle x,y \rangle} s_x s_y$ beeing forced to 1 if $\sigma = 1$.

We observe that:

$$\prod_{\substack{\sigma=0 \\ J s_x^1 s_y^1 = -1}} e^{-\beta} \times \prod_{\substack{\sigma=0 \\ J s_x^1 s_y^1 = 1}} e^{-\beta} = \prod_{\sigma=0} e^{-\beta} = k$$

k depends only on the link configuration $\{\sigma\}$

$$\prod_{\substack{\sigma=0 \\ J s_x^1 s_y^1 = -1}} e^{-\beta} = k \prod_{\substack{\sigma=0 \\ J s_x^1 s_y^1 = 1}} e^{\beta}$$

$$\frac{e^{-\beta \mathcal{H}(\{s_1\})}}{e^{-\beta \mathcal{H}(\{s_0\})}} = \frac{\prod_{J s_x^1 s_y^1 = +1}^{\sigma=0} e^{2\beta}}{\prod_{J s_x^0 s_y^0 = +1}^{\sigma=0} e^{2\beta}} = \frac{\prod_{J s_x^0 s_y^0 = +1}^{\sigma=0} e^{-2\beta}}{\prod_{J s_x^1 s_y^1 = +1}^{\sigma=0} e^{-2\beta}}$$

We arrived at:

$$\frac{P(\{s_0, \sigma_0\} \rightarrow \{s_1, \sigma\})}{P(\{s_1, \sigma_0\} \rightarrow \{s_0, \sigma\})} = \frac{e^{-\beta \mathcal{H}(\{s_1\})}}{e^{-\beta \mathcal{H}(\{s_0\})}}$$

$$P(\{s_0\} \rightarrow \{s_1\}) = \sum_{\{\sigma\}, \{\sigma_0\}} P(\{s_0, \sigma_0\} \rightarrow \{s_1, \sigma\})$$

We thus obtain the detailed balance:

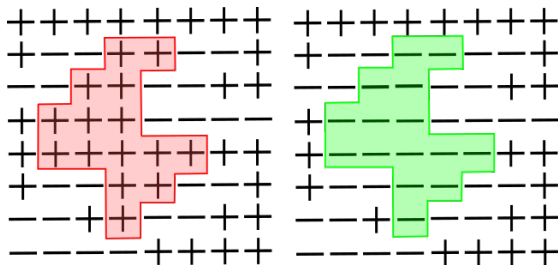
$$\frac{P(\{s_0\} \rightarrow \{s_1\})}{P(\{s_1\} \rightarrow \{s_0\})} = \frac{e^{-\beta \mathcal{H}(\{s_1\})}}{e^{-\beta \mathcal{H}(\{s_0\})}}$$

- The only requirement on the update step is that $P(\{s_1\}|\{\sigma, s_0\}) = P(\{s_0\}|\{\sigma, s_1\})$.
- We can also update the coupling constants $J_{\langle x,y \rangle}$ as long as $J_{\langle x,y \rangle} s_x s_y = 1$ where $\sigma = 1$.

These two observations give rise to two key modifications of the SW algorithm: the Wolff algorithm and the boundary flip.

Single cluster update (Wolff):

In the Wolff algorithm we choose at random one site of the lattice and flip the cluster it belongs to. The probability of going from s_1 to s_0 and viceversa is obviously the same, being the probability of choosing the right cluster. Notice that $P(\{s\}|\{\sigma\}) \neq \text{const.}$ Practically we build only **one cluster** starting from a seed.



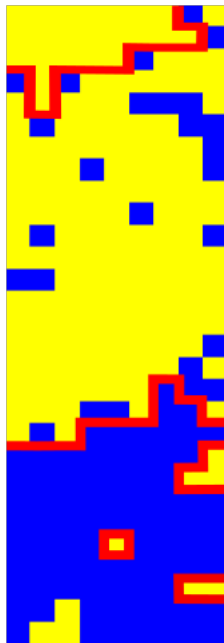
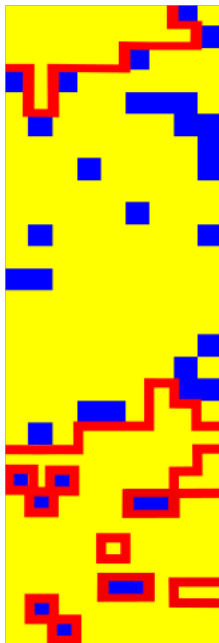
Boundary flip algorithm:

$J = 1$ in the bulk, but the coupling between the top and the bottom is now a dynamical variable to simulate:

$\{s, \sigma, J_{0,T-1} = \pm 1\}$ is an element of the ensemble.

A link $\sigma_{\langle 0, T-1 \rangle} = 1$ demands $J_{0,T-1} s_0 s_{T-1} = 1$. Thus we can flip $J_{0,T-1}$, s_0 and all the spins connected to s_0 via some chain of links in the bulk obtaining a configuration compatible with $\{\sigma_{\langle x,y \rangle}\}$. This for all the spins on the bottom boundary.

Boundary
condition
update.
Section of a
 $7 \times 7 \times 21$
lattice at
temperature
 $\beta = 0.250$

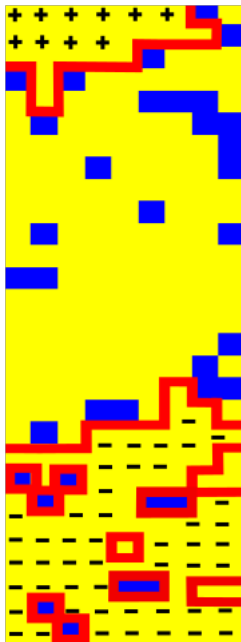


The boundary flip generates an interface between phases!

We must check this update can be done without violating the constraints imposed by the bulk links ($\sigma_{\langle x,y \rangle} = 1$ implies $s_x = s_y$).

Check all the clusters that contain sites of the lower surface and flip the boundary condition only if it can be done consistently in all the lattice.

Introduce extra
variables $c_x = \pm 1$
that get
propagated in the
bulk during the
costruction of the
cluster but change
sign when crossing
the boundary.



We do N total steps alternating the Wolff algorithm with the boundary flip.

After each update we count if the current configuration has ferromagnetic or antiferromagnetic coupling between top and bottom.

$$\langle \delta_{J=-1} \rangle = \frac{\# \text{Antiferromagnetic}}{N}$$

$$\langle \delta_{J=+1} \rangle = \frac{\# \text{Ferromagnetic}}{N}$$

Notes on the implementation

C++ for the Montecarlo and Jackknife algorithms, Python for data analysis, fits and plots.

The hot function of the simulation generates a cluster starting in a given position and exploring the neighbouring links. If a link is chosen to be $\sigma_{\langle x,y \rangle} = 1$ then the adjacent site is included in the cluster and the procedure is repeated. The extra variable c_x is also propagated.

```
stack<site> stack
stack.push(seed)
while(!stack.empty())
    site current = stack.top()
    if(cluster[current] is incostintent) flag = 1;
    else if(cluster[current] == 0)
        cluster[current] = cluster[old];
        for(d = 0; d < 3; d++)
            for(a = -1; a < 2; a = a + 2)
                next = current + a
                Check if we are on the boundary
                if(cluster[next] == 0)
                    if(p > 0 and random < p) stack.push(next)
return flag
```

- The simulations were executed with $N = 10^5, 10^6$ steps of the Markov chain.
- The first 5% of the Markov chain is ignored to avoid non-thermalized configurations.
- The Jackknife is executed for blocks of size starting from 10000 to 10900 with step of 30 events.

Estimation of the errors and autocorrelation

Fit of the free energy - theory

Fit of the free energy - results

Fit of the critical behaviour

(Conclusions?)