

Swendsen-Wang beats Heat-bath

Mario Ullrich

Friedrich Schiller University Jena

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Overview

- 1 Introduction
 - Potts model
 - Random cluster model
- 2 Results
 - Main result
 - New algorithm

The Problem

Problem:

Can one construct a rapidly mixing
Markov chain for the Potts model
at all temperatures?

The Potts Model

For $q \in \mathbb{N}$ and a (finite) graph $G = (V, E)$, $N := |V|$, the q -state Potts model on G is defined as the set of possible configurations $\Omega_P = \{1, \dots, q\}^V$ together with the probability measure

$$\pi_\beta(\sigma) := \frac{1}{Z(G, \beta, q)} \exp \left\{ \beta \cdot \# \left\{ \{u, v\} \in E : \sigma(u) = \sigma(v) \right\} \right\}$$

for $\sigma \in \Omega_P$, where Z is the normalization constant and $\beta \geq 0$ is the inverse temperature.

The Dynamics I

The most studied Markov chain is the *heat bath dynamics*. This is the Markov chain with transition probabilities

$$P_{\text{HB}}(\sigma, \sigma^{v,k}) := P_{\text{HB},\beta,q}^G(\sigma, \sigma^{v,k}) = \frac{1}{N} \frac{\pi_\beta(\sigma^{v,k})}{\sum_{l=1}^q \pi_\beta(\sigma^{v,l})},$$

where $\sigma^{v,k}(v) = k$ and $\sigma^{v,k}(u) = \sigma(u)$, $u \neq v$.

Rapid Mixing

For the transition matrix P of a Markov chain on state space Ω with stationary distribution π , we define the operator

$P : L_2(\pi) \rightarrow L_2(\pi)$ by

$$Pf(x) := \sum_{y \in \Omega} P(x, y) f(y).$$

If $1 = \xi_1 \geq \xi_2 \geq \dots \geq \xi_{|\Omega|} \geq -1$ are the (real) eigenvalues of the operator P , we define the *spectral gap* of the Markov chain by

$$\lambda(P) = 1 - \max\{\xi_2, |\xi_{|\Omega|}|\}.$$

Rapid Mixing

Let $\{P_n\}_{n \in \mathbb{N}}$ be a *family of Markov chains* with corresponding state spaces Ω_n . Then we call the Markov chains *rapidly mixing* (for $\{\Omega_n\}$), if

$$\lambda(P_n)^{-1} = \mathcal{O}\left((\log |\Omega_n|)^C\right) \quad \text{for some } C \geq 0 \text{ and all } n \in \mathbb{N}.$$

In "our" case let $\{G_n\}_{n \in \mathbb{N}}$ be a *family of graphs*, then $\log |\Omega_n| = \log q \cdot |V_{G_n}|$.

Mixing for heat bath

An example of physical interest:

The two-dimensional square lattice $\mathbb{L}_N = (V, E)$ with $V = \{1, \dots, L\}^2$, $N = L^2$, and $E = \{(v, w) \in V^2 : |v - w| = 1\}$.

It is well known that $P_{\text{HB}} = P_{\text{HB}, \beta, q}^{\mathbb{L}_N}$ satisfies

- $\lambda(P_{\text{HB}})^{-1} = \mathcal{O}(N)$, if $\beta < \beta_c(q) := \ln(1 + \sqrt{q})$.
- $\lambda(P_{\text{HB}})^{-1} = e^{\Omega(N)}$, if $\beta > \beta_c(q)$.
- $\lambda(P_{\text{HB}})^{-1} = \mathcal{O}(N^C)$, if $q = 2$ and $\beta = \beta_c(2)$, for some $C > 0$.

(Martinelli et al. ('94), Cesi et al. ('96), Lubetzky & Sly (2010), Beffara & Duminil-Copin (2010))

Random cluster model

The *random cluster model* on $G = (V, E)$ has the state space $\Omega_{\text{RC}} = \{\omega \subseteq E\}$. So $\omega \in \Omega_{\text{RC}}$ induces a subgraph (V, ω) of (V, E) .

The probability measure on Ω_{RC} is given by

$$\mu_p(\omega) = \frac{1}{Z} p^{|\omega|} (1-p)^{|E|-|\omega|} q^{C(\omega)},$$

where $C(\omega)$ is the number of connected components in (V, ω) .

Connection of the models

In the case $p = 1 - e^{-\beta}$, the Potts model and the random cluster model are equivalent in the following sense.

We can define random mappings $T : \Omega_P \mapsto \Omega_{RC}$ and $T^* : \Omega_{RC} \mapsto \Omega_P$ such that

$$X \sim \pi_\beta \implies T(X) \sim \mu_p$$

and

$$Y \sim \mu_p \implies T^*(Y) \sim \pi_\beta.$$

We consider the Markov chain

$$\sigma_{t+1} = T^* \circ T(\sigma_t).$$

The Dynamics II

The most widely *used* algorithm is the *Swendsen-Wang dynamics*.
For $\sigma \in \Omega_P$ let

$$E(\sigma) := \{\{u, v\} \in E : \sigma(u) = \sigma(v)\}.$$

The chain performs the following steps:

- 1 Given a Potts configuration $\sigma_t \in \Omega_P$ on G , delete each edge of $E(\sigma_t)$ independently with probability $1 - p = e^{-\beta}$. This gives $\omega \in \Omega_{RC}$.
- 2 Assign a random color independently to each connected component of (V, ω) . Vertices of the same component get the same color. This gives $\sigma_{t+1} \in \Omega_P$.

Mixing for Swendsen-Wang

- Empirically, SW is much faster than HB, but there are not many results on the mixing properties of SW.
- Positive results are only known for special classes of graphs (trees, cycles, complete graph etc.) or for high enough temperatures.
- A negative result: Slow mixing on \mathbb{Z}^d , $d \geq 2$, at $\beta \approx \beta_c(q)$ for q large enough. (Borgs, Chayes & Tetali (2010))

Main result

Theorem (U 2011)

Suppose that P (resp. P_{HB}) is the transition matrix of the Swendsen-Wang (resp. heat-bath) dynamics, which is reversible with respect to $\pi_{\beta,q}^G$. Then

$$\lambda(P) \geq c_{\text{SW}} \lambda(P_{\text{HB}}),$$

where

$$c_{\text{SW}} = c_{\text{SW}}(G, \beta, q) := \frac{1}{2q^2} \left(q e^{2\beta} \right)^{-4\Delta},$$

where Δ is the maximal degree of G .

Corollary

With this theorem we get new and old results on SW (up to a factor N). Especially, we get rapid mixing for the two-dimensional square lattice \mathbb{L}_N :

Corollary (Square lattice \mathbb{L}_N)

Let $P = P_{p,q}^{\mathbb{L}_N}$. Then

- $\lambda(P)^{-1} = \mathcal{O}(N)$, if $\beta < \beta_c(q) := \ln(1 + \sqrt{q})$.
- $\lambda(P)^{-1} = \mathcal{O}(N^C)$, if $q = 2$ and $\beta = \beta_c(2)$.

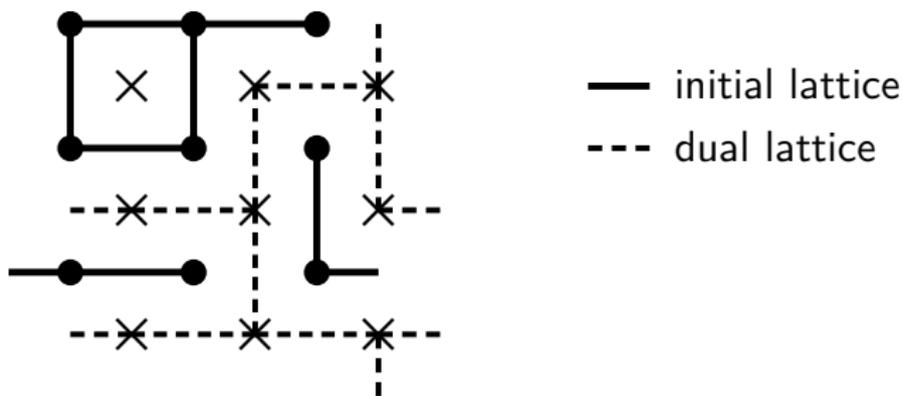
Comments

- The bounds are (probably) off by a factor N , because we compare a local and a highly non-local algorithm.
- One may expect rapid mixing also at low temperatures. We were not able to prove this, but to modify the algorithm.

Dual Lattices

Let $G = (V, E)$ be a finite, **planar** graph and $G_D = (V_D, E_D)$ its dual graph. Then, to each RC configuration $\omega \in \Omega_{RC}$ there corresponds the dual configuration $\omega_D \subseteq E_D$, given by

$$e_D \in \omega_D \iff e \notin \omega.$$



Duality

It is easy to obtain (using Euler's polyhedron formula)

$$\mu_{p,q}^G(\omega) = \mu_{p^*,q}^{G_D}(\omega_D),$$

where the dual parameter p^* satisfies

$$\frac{p^*}{1-p^*} = \frac{q(1-p)}{p}.$$

The algorithm

Let $\tilde{P} = \tilde{P}_{p,q}^G$ be the SW algorithm on the random cluster model.

The dynamics performs the following steps:

- 1 Given a Potts configuration σ_t on G , generate a random cluster state $\omega \subset E_G$ by $\omega = T(\sigma_t)$.
- 2 Make one step of the Swendsen-Wang dynamics $\tilde{P}_{p^*,q}^{G_D}$ starting at $\omega_D \subset E_{G_D}$ to get a random cluster state $\tilde{\omega}_D \subset E_{G_D}$.
- 3 Generate σ_{t+1} by $T^*(\tilde{\omega})$.

Denote by M the transition matrix of this Markov chain.

Result

Proposition (U 2011)

Let $P_{p,q}^G$ be the Swendsen-Wang dynamics on a planar graph G , which is reversible with respect to $\pi_{\beta,q}^G$, and M as above. Then

$$\lambda(M) \geq \max\left\{\lambda(P_{p,q}^G), \lambda(P_{p^*,q}^{G_D})\right\}.$$

We obtain for the square lattice \mathbb{L}_N :

- $\lambda(M)^{-1} = \mathcal{O}(N)$, if $\beta \neq \beta_c(q) := \ln(1 + \sqrt{q})$.
- $\lambda(M)^{-1} = \mathcal{O}(N^C)$, if $q = 2$ and $\beta = \beta_c(2)$.

Partial answers

Problem:

Can one construct a rapidly mixing
Markov chain for the Potts model
at all temperatures?

For the square lattice \mathbb{L}_N :

- Yes, for $q = 2$.
- Yes, for $q > 2$ and $\beta \neq \beta_c(q)$.
- Probably not for $\beta = \beta_c(q)$ and q large enough.

Thank you!