Generalized cluster algorithms for Potts lattice gauge theory [arxiv:2507.13503]

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GEORGE INST

A crash course in probability

The q-state, (i-1)-dim Potts lattice gauge theory (PLGT) on a d-dim cubical complex X [1–3] is the family of probability distributions $u_{eta,q}(f) \propto$ $e^{-\beta \mathbf{H}(f)}$ over (i-1)-cochains f where

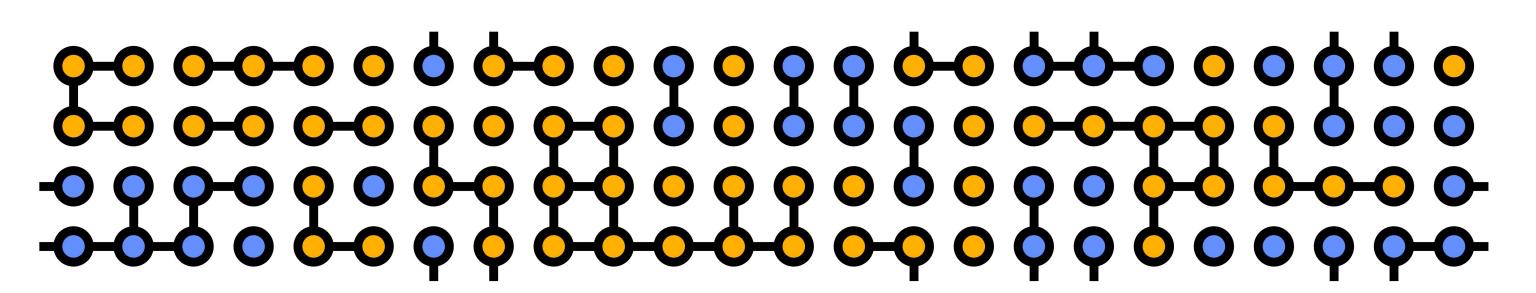
$$\mathbf{H}(f) = \sum_{x \in X^i} \mathbf{1} \left((\delta^{i-1} f)(x) = 0 \right).$$

The i-dim plaquette random-cluster model (PRCM) $\mu_{p,q}$ on X [4, 5] is the random subcomplex $P \subseteq X^i$ with distribution

$$\mu_{p,q}(P) \propto p^{|P|} (1-p)^{|X^i|-|P|} |H^{i-1}(P; \mathbb{Z}/q\mathbb{Z})|.$$

When i=1, PLGT is the q-state Potts model and PRCM is the *q*-state random-cluster model.

For some intuition, *Figure 1* shows a low-dim example of PLGT and PRCM.



when a subcomplex P and cochain f are compatible...

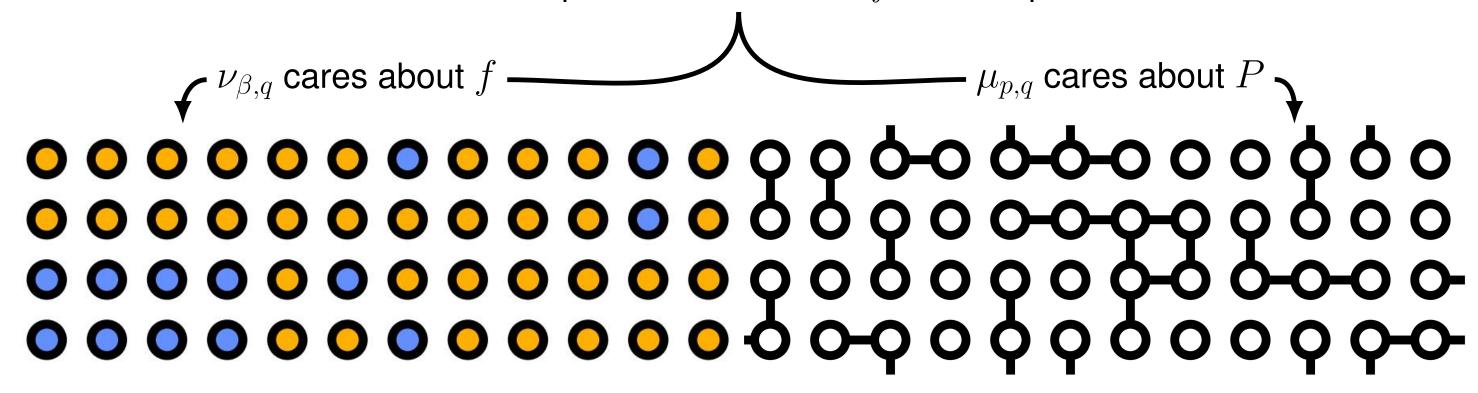


Figure 1. When $i=1, X^i$ is a graph, so f is a vertex coloring and P is an edge-subgraph. Since q=2, PLGT is the *Ising model*.

A cochain f and subcomplex P are **compatible** when f is a cocycle on P. The **Edwards-Sokal coupling** K is a joint distribution over compatible pairs (f, P) [4–6].

If (f, P) is a compatible pair, $\mathcal{K}(-\mid P)$ is the uniform distribution over cocycles $Z^{i-1}(P; \mathbb{Z}/q\mathbb{Z})$, and $\mathcal{K}(-\mid f)$ is independent percolation over f-compatible i-cells.

Later, we use the coupling to build efficient sampling algorithms.

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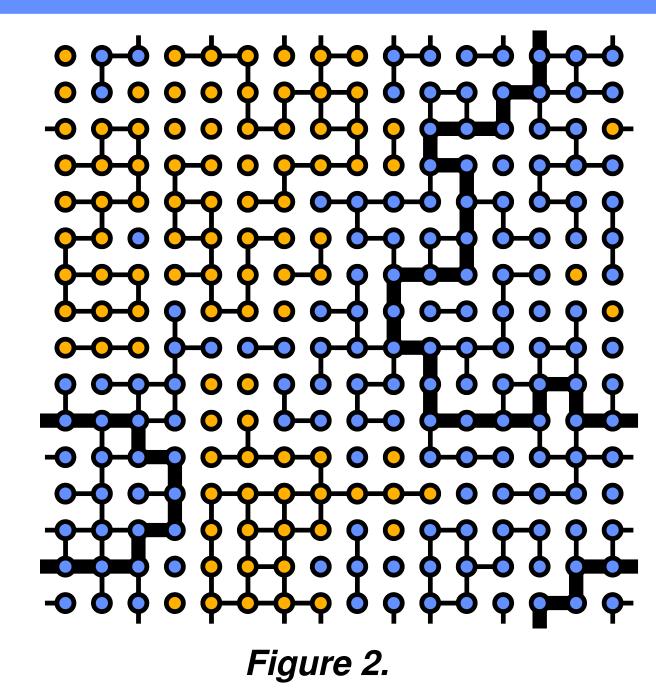
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Detecting phase transitions with giant cycles

Giant i-cycles are the nontrivial elements of $H_i(\mathbb{T}^d,\mathbb{Z}/q\mathbb{Z})$. Homological percolation is the event that $PH_i(P, \mathbb{Z}/q\mathbb{Z})$ is nontrivial [7, 8]. For i=1, the phase transition in PLGT occurs when giant 1-cycles appear in the PRCM (like in *Figure 2*).

When d=2i and q is an odd prime, homological percolation in the PRCM happens at the self-dual point $p_{\rm sd}(q) = \sqrt{q}/(1+\sqrt{q})$ of PLGT, which physicists believe is PLGT's critical point. We conjecture that PLGT's phase transition occurs exactly when homological percolation does in the PRCM [3].

We can use the coupling K, persistent homology, and linear algebra to generalize two algorithms for simulating these systems on *scale-N* cubical d-torus \mathbb{T}_N^d .



Algorithms, performance, and future work

The plaquette invaded-cluster (PIC) [9, 10] and plaquette Swendsen-Wang (PSW) [10, 11] algorithms sample PLGT by iteratively updating a cochain f, each passing through a f-compatible subcomplex in a different way.

Empirically, Algorithms 1 and 2 blow past standard algorithms for sampling PLGT and PRCM, even on extremely large systems [10]. See Figure 4.

PSW provably targets PLGT. PIC estimates homological percolation thresholds for \mathbb{T}_N^d [10].

We expect PIC to converge to PLGT as $N \longrightarrow \infty$, according to the scaling behavior in [3].

Algorithm 1 (PIC update routine).

- **1.** Let C be the set of f-compatible i-cells in Xand $P:=X^{i-1}$.
- **2.** Pick uniform random i-cells in C and add them to P until $PH_i(P, \mathbb{Z}/q\mathbb{Z}) \neq \emptyset$.
- 3. Sample and return $f' \sim \mathfrak{K}(-\mid P)$. 3 4

Algorithm 2 (PSW update routine).

- **1.** Let C be the set of f-compatible i-cells in Xand $P:=X^{i-1}$.
- **2.** Add each i-cell in C to P with probability $p = 1 - e^{-\beta}$. 2
- 3. Sample and return $f' \sim \mathfrak{K}(-\mid P)$. 3 4

Subway markers (e.g. 2) match the adjacent step to its stop on the core loop in Figure 3.

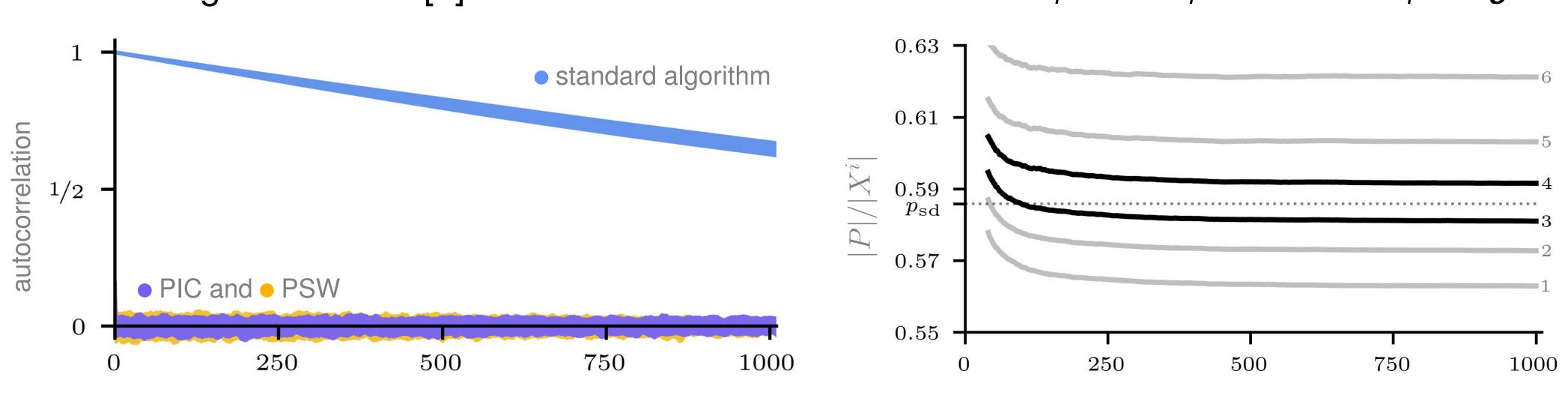


Figure 4. (Left) autocorrelation decay of $\mathbf{H}(f)$, comparing PIC and PSW to a standard algorithm called *Glauber dynamics* on \mathbb{T}_{10}^4 by measuring the similarity of (f_t, P_t) to the initial pair (f_0, P_0) . (Right) expected thresholds for finding k independent cycles on \mathbb{T}^4_{10} .

As the system size increases, we believe PIC will overtake PSW in convergence rate when the system is at or near criticality.

Both Algorithms are implemented in ATEAMS, our companion software [12]. Depending on q, we use PHAT or a modified twist-reduce for persistence, and LinBox for sparse linear algebra [12–14]. Currently, Π_{40}^4 is the largest system we have simulated successfully.

Future work includes: running experiments on larger systems; implementing a probabilistic algorithm for persistence; further optimizing ATEAMS; estimating critical