

A worm-inspired algorithm for the simulation of Abelian gauge theories

Tomasz Korzec*

Institut für Physik, Humboldt Universität zu Berlin, Newtonstrasse 15, 12489 Berlin, Germany E-mail: korzec@physik.hu-berlin.de

Ulli Wolff

Institut für Physik, Humboldt Universität zu Berlin, Newtonstrasse 15, 12489 Berlin, Germany E-mail: uwolff@physik.hu-berlin.de

We present an algorithm in which the all-order strong coupling expansion of the Abelian U(1) gauge theory with Wilson plaquette action is sampled. In addition to the vacuum closed surface graphs of the partition function we propose to also allow for a class of defects (boundaries) related to Wilson loops in the ensemble. The efficiency of our scheme in estimating various observables is compared to a standard Metropolis algorithm.

The XXVIII International Symposium on Lattice Field Theory, Lattice2010 June 14-19, 2010 Villasimius, Italy

*St	oeal	ker.
-----	------	------

1. Worm algorithms

Monte-Carlo simulation algorithms based on all-order strong coupling or hopping parameter expansions have been known for a long time [1]. Yet only the relatively recent ideas of Prokof'ev and Svistunov [2] opened the way to the formulation of highly efficient algorithms. Instead of sampling the partition function of a model the new class of algorithms samples an enlarged ensemble that also contains the two-point-functions at all possible separations. The "worm algorithms" that can be formulated in the loop gas representation of such an ensemble typically show hardly any critical slowing down. In addition the non-standard formulation often makes improved estimators for key observables possible. The basic idea behind worm algorithms is very general and can be applied to a large class of statistical models. Efficient algorithms have been developed for the one (real or complex) component ϕ^4 theory and tested in its Gaussian and Ising or XY-model limits as well as at some intermediate values of the coupling [2, 3, 4, 5, 6, 7]. More complicated systems like nonlinear $O(N) - \sigma$ models [8], CP(N-1) models [9] or two dimensional fermionic systems [10, 11] have been successfully simulated as well. Some of the recent development has been summarized during this conference [12].

In the following we make a first attempt to generalize the idea behind worm algorithms to the case of gauge theories. We formulate an algorithm that samples the (generalized) partition function of an Abelian U(1) gauge theory.

2. Strong coupling expansion

We start from a Wilson plaquette gauge action

$$S[U] = -\beta \sum_{x,\mu < \nu} \text{Re} \left[U(x,\mu)U(x+\hat{\mu},\nu)U^{-1}(x+\hat{\nu},\mu)U^{-1}(x,\nu) \right], \qquad (2.1)$$

where x is a site on a D-dimensional hypercubic periodic lattice of extent L_{μ} in direction $\mu = 1 \dots D$ and $U(x, \mu) \in U(1)$ denotes the gauge field on the link that connects site x with site $x + \hat{\mu}$. We use lattice units throughout.

Observables in this model are products of link variables

$$U^{j} = \prod_{x\mu} U(x,\mu)^{j(x,\mu)}, \qquad (2.2)$$

where we have introduced an integer valued external field $j(x, \mu) \equiv j_{\mu}(x) \in \mathbb{Z}$. One is interested in their expectation values

$$\langle U^j \rangle = \frac{Z[j]}{Z[0]},\tag{2.3}$$

$$Z[j] = \int DU \ e^{-S[U]} U^j, \qquad (2.4)$$

where DU denotes the invariant measure on U(1) on all links. Gauge and center invariance of the action causes expectation values to vanish unless the external field satisfies

$$\partial_{\mu}^* j_{\mu}(x) = 0, \quad \text{and}$$
 (2.5)

$$\sum_{x} j_{\mu}(x) = 0, (2.6)$$

where ∂^* denotes the backward nearest neighbor difference.

On each plaquette one can expand the exponential

$$e^{\beta \text{Re}(U)} = \sum_{m,n=0}^{\infty} \left(\frac{\beta}{2}\right)^{m+n} \frac{U^{n-m}}{m!n!} = \sum_{n=-\infty}^{+\infty} I_n(\beta) U^n.$$
 (2.7)

The summation variables $n(x; \mu, \nu) \equiv n_{\mu\nu}(x)$ can be defined to be antisymmetric in μ, ν . At this stage the original group integrals can be carried out leaving behind constraints. The partition function takes the form

$$Z[j] = \sum_{\{n\}} \left(\prod_{x,\mu < \nu} I_{n_{\mu\nu}(x)}(\beta) \right) \delta[\partial^* n - j], \qquad (2.8)$$

with the abbreviation for the constraints

$$\delta[\partial^* n - j] = \prod_{x\mu} \delta_{\partial_v^* n_{\nu\mu}(x), j_\mu(x)}. \tag{2.9}$$

The representation eq. (2.8) has been used as a starting point for Monte-Carlo simulations in [13] and more recently in [14]. The authors of [14] find that a variant of their algorithm, that is based on an expansion of Z[0] of the Z_2 gauge theory only, suffers from critical slowing down with a dynamical exponent similar to that of a local algorithm in the standard formulation of the model.

Inspired by worm-algorithms we pursue a different direction. In spin systems the crucial algorithmic idea was to consider an enlarged system that allows for defects related to two-point functions. Similarly in our case we consider the enlarged ensemble

$$\mathscr{Z} = \sum_{\{j\}} \rho^{-1}[j] Z[j]. \tag{2.10}$$

The sum is over all possible external fields $j_{\mu}(x)$ and, as in [4], each contribution is weighted by some non-negative weight ρ . \mathscr{Z} contains now also graphs with boundaries rather than closed surfaces only. The partition function can be written in its final form

$$\mathscr{Z} = \sum_{\{n\}} \left(\prod_{x,\mu < \nu} I_{n_{\mu\nu}(x)}(\beta) \right) \rho^{-1} [\partial^* n]. \tag{2.11}$$

Expectation values with respect to this ensemble are formed in the usual way

$$\langle \langle \mathscr{O}[n] \rangle \rangle = \frac{1}{\mathscr{Z}} \sum_{\{n\}} \mathscr{O}[n] \left(\prod_{x, \mu < \nu} I_{n_{\mu\nu}(x)}(\beta) \right) \rho^{-1} [\partial^* n], \qquad (2.12)$$

and for later convenience and in analogy to ref. [8] we also define an expectation value with respect to the vacuum ensemble by

$$\langle \langle \mathscr{O}[n] \rangle \rangle_0 = \frac{\langle \langle \mathscr{O}[n] \delta[\partial^* n] \rangle \rangle}{\langle \langle \delta[\partial^* n] \rangle \rangle}. \tag{2.13}$$

One can relate observables in the enlarged ensemble to those in the original one. One way to measure the expectation value of an arbitrary non-vanishing correlator, given by some $j_{\mu}(x)$, is to find some background plaquette-field $k_{\mu\nu}(x)$ that solves the constraint $\partial^* k = j$. The observable

$$\mathscr{O}[n] = \prod_{x,\mu < \nu} \frac{I_{n_{\mu\nu}(x) + k_{\mu\nu}(x)}(\beta)}{I_{n_{\mu\nu}(x)}(\beta)}$$
(2.14)

is then an estimator for the desired correlator

$$\langle U^j \rangle = \langle \langle \mathscr{O}[n] \rangle \rangle_0. \tag{2.15}$$

In practice k will be zero almost everywhere. For rectangular loops the minimal number of factors differing from one necessary in the product eq. (2.14) equals their area.

3. Monte-Carlo updates

We use the freedom in choosing the weight function ρ^{-1} to restrict the possible external fields or boundaries to the subset containing only one non-intersecting loop with zero winding number with respect to all torus directions. Such a loop can be completely characterized by the ordered cyclic set of sites through which it passes

$$\mathcal{W} = \{x_1, x_2, \dots, x_{P(\mathcal{W})}\}, \quad x_{P(\mathcal{W})+1} = x_1.$$
(3.1)

Consecutive sites are nearest neighbors $x_{i+1} = x_i + \hat{\mu}_i$. If we extend the possible unit vectors to include negative directions: $\widehat{-\mu} \equiv -\hat{\mu}$, plaquettes can be labeled in several equivalent ways

$$n_{\mu\nu}(x) = -n_{-\mu\nu}(x+\hat{\mu}) = -n_{\mu-\nu}(x+\hat{\nu}) = n_{-\mu-\nu}(x+\hat{\mu}+\hat{\nu}). \tag{3.2}$$

To stochastically sample the sum eq. (2.11) we perform a sequence of local updates of which each by itself satisfies detailed balance. The updates consist of a proposal $\mathcal{W} \to \mathcal{W}'$ followed by a Metropolis accept/reject step.

3.1 Flips

If the perimeter P(W) is larger than four we pick a random site x_i on the loop and form $y = x_{i+1} - x_i + x_{i-1}$. If $y \in W$ the old configuration is kept, otherwise the proposal is to form W' by replacing $x_i \to y$ and to adjust the plaquette field accordingly

$$n_{-\mu_{i-1}\mu_i}(x_i) \to n_{-\mu_{i-1}\mu_i}(x_i) + 1.$$
 (3.3)

Such a proposal is accepted with probability $min(1, q_{flip})$ with

$$q_{\text{flip}} = \frac{I_{n_{-\mu_{i-1}\mu_{i}}(x_{i})+1}(\beta)}{I_{n_{-\mu_{i-1}\mu_{i}}(x_{i})}(\beta)} \frac{\rho(\mathscr{W})}{\rho(\mathscr{W}')}.$$
(3.4)

This update changes the shape of the loop without altering its perimeter. It is displayed in fig. 1.

3.2 Shifts

A site x_i and a direction \hat{v} orthogonal to $\hat{\mu}_i$ is chosen randomly. Two sites, $y = x_i + \hat{\mu}$ and $y' = x_{i+1} + \hat{v}$ are constructed. A proposal is made in either of two cases

1. If neither $y \in \mathcal{W}$ nor $y' \in \mathcal{W}$, it is proposed to extend the loop by two sites

$$\mathcal{W}' = \{x_1, \dots, x_i, y, y', x_{i+1}, \dots\},\tag{3.5}$$

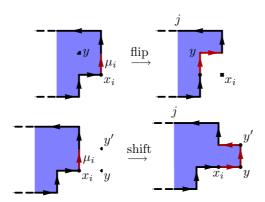


Figure 1: Flip (upper part) and shift (lower part) updates. The plaquette field in the shaded area differs by one unit with respect to the non-shaded areas.

together with the necessary change in the plaquette field: $n_{\nu\mu_i}(x_i) \to n_{\nu\mu_i}(x_i) + 1$. In analogy to the flips we accept with

$$q_{\text{shift}} = \frac{I_{n_{\nu\mu_i}(x_i)+1}(\beta)}{I_{n_{\nu\mu_i}(x_i)}(\beta)} \frac{\rho(\mathcal{W})}{\rho(\mathcal{W}')}.$$
(3.6)

2. If $y = x_{i-1}$ and $y' = x_{i+2}$ the reverse move is proposed, i.e. x_i and x_{i+1} are proposed to be removed from the loop. The acceptance probability is again $min(1, q_{shift})$.

Otherwise no move is made.

3.3 Non-local updates

With free boundary conditions the alternation of the two updates proposed so far would be sufficient to guarantee ergodicity. On a torus however, there exist contributions to the partition function that are not yet sampled. This is fixed by another update:

A plaquette $x, \mu, \nu \neq \mu$ is chosen randomly. The subsets of sites which belong to the two dimensional plane through x spanned by $\hat{\mu}$ and $\hat{\nu}$ is denoted by Γ . The proposal is to change the plaquette field according to $\hat{\nu}$

for all
$$x \in \Gamma$$
: $n_{\mu\nu}(x) \to n_{\mu\nu}(x) + 1$ (3.7)

and it is accepted with

$$q_{\text{plane}} = \prod_{x \in \Gamma} \frac{I_{n_{\mu\nu}(x)+1}(\beta)}{I_{n_{\mu\nu}(x)}(\beta)}.$$
 (3.8)

The acceptance rate for this proposal is tiny unless the volume is small.

The last update that we have implemented is proposed whenever \mathcal{W} is planar, or alternatively whenever it has exactly 4 corners. The proposal is to shift the whole loop by one unit in a direction perpendicular to its plane. This update can be built up from the elementary shifts and flips, but it turns out to reduce autocorrelation times considerably if it is introduced as a separate update step.

One iteration of our algorithm consists of order of volume many shift, flip and planar-loop-shift updates followed by one plane update. The costs are comparable to those of a Metropolis sweep in the standard formulation, i.e. O(volume).

¹The inverse move is realized by swapping $\mu \leftrightarrow \nu$.

4. Performance of the algorithm

Abelian gauge theory is particularly well understood in three dimensions. For the model with Villain action, which is believed to lie in the same universality class, the existence of a mass gap m and confinement at every value of β have been proven [15]. The continuum limit at fixed m is believed to describe free massive bosons. A recent numerical work with the Wilson action [16] showed that the mass gap is indeed well described by

$$mL = 5.23(11) \frac{L}{a} \sqrt{8\pi^2 \beta} \exp\left[-0.2527\pi^2 \beta\right].$$
 (4.1)

To test our algorithm we use eq. (4.1) to keep the physical volume constant at $mL \approx 6$ while increasing the resolution $L/a \in \{8,16,24,32,40\}$. As weight function we take an exponential that depends on the loop perimeter $\rho[\mathcal{W}] = \exp(\theta(P[\mathcal{W}] - 2))$, and tune the "loop tension" θ to an algorithmically close to optimal value. More precisely, the cost to estimate a $L/4 \times L/4$ Wilson loop to a given precision is minimalized. We measure expectation values of rectangular Wilson loops as well as of plaquette-plaquette correlators. The autocorrelation times and a cost-indicator (CPU-time \times relative error squared divided by volume) of the average plaquette, the string tension and an effective mass from the plaquette-plaquette correlator at separation L/4 are monitored. The same calculations are repeated with a standard Metropolis algorithm, in which the proposals are to change the gauge field on a link $e^{i\phi} \rightarrow e^{i(\phi+\Delta\phi)}$. In this case the number of sweeps per measurement as well as the size of the interval from which $\Delta\phi$ is drawn are kept at their optimal values. The optimal number of sweeps per measurement is determined only on the 16^3 lattice and kept at the same value (i.e. 15) for the others. The size of the interval is determined on each lattice separately. Expectation values obtained with both algorithms are consistent with each other and a comparison of autocorrelation times and costs is shown in Fig. 2.

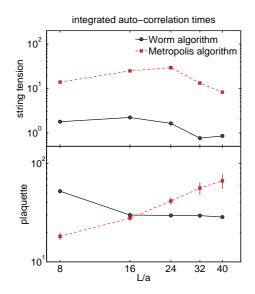
5. Conclusions

We have extended the concept of worm algorithms to Abelian gauge theories. Instead of a loop gas we deal with a surface ensemble. The worm is replaced by an open surface and a Wilson loop plays the role analogous to the worm's head and tail. All standard observables can be estimated in this model. In first numerical tests in three dimensions no significant critical slowing down could be observed, but larger correlation lengths will be necessary to make a definite statement. On the presently available lattices also a standard Metropolis algorithm performs relatively well.

Abelian gauge theories with different actions, like the Villain model or Wegner's Z_2 model can presumably be treated in the same way. Whether matter fields can be incorporated into the algorithm or whether an extension to non Abelian models is feasible remains to be investigated.

References

- [1] B. Berg and D. Förster *Phys. Lett.* **B106** (1981) 323.
- [2] N. Prokof'ev and B. Svistunov Phys. Rev. Lett. 87 (2001) 160601.
- [3] Y. Deng, T. M. Garoni and A. D. Sokal *Phys. Rev. Lett.* **99** (2007) 110601 [arXiv:cond-mat/0703787].



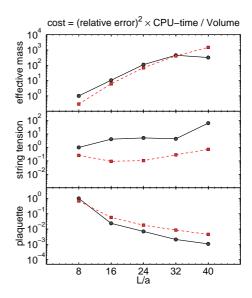


Figure 2: A comparison of the new algorithm with a standard Metropolis one. The left panel shows integrated autocorrelation times of two observables in units of sweeps or iterations. The right panel compares the costs. The autocorrelation times of the effective mass are consistent with 0.5 for this set of lattices and are not plotted.

- [4] U. Wolff Nucl. Phys. **B810** (2009) 491–502 [arXiv:0808.3934].
- [5] U. Wolff *Phys. Rev.* **D79** (2009) 105002 [arXiv:0902.3100].
- [6] W. Janke, T. Neuhaus and A. M. J. Schakel *Nucl. Phys.* **B829** (2010) 573–599 [arXiv:0910.5231].
- [7] I. Vierhaus, Simulation of ϕ^4 Theory in the Strong Coupling Expansion beyond the Ising Limit, Diploma thesis, Humboldt Universität zu Berlin, 2010, and T. Korzec, I. Vierhaus, U. Wolff, in preparation.
- [8] U. Wolff Nucl. Phys. **B824** (2010) 254–272 [arXiv:0908.0284].
- [9] U. Wolff Nucl. Phys. **B832** (2010) 520–537 [arXiv:1001.2231].
- [10] U. Wolff Nucl. Phys. **B814** (2009) 549-572 [arXiv:0812.0677].
- [11] U. Wenger *Phys. Rev.* **D80** (2009) 071503 [arXiv:0812.3565].
- [12] U. Wolff PoS(Lattice 2010)020(2010)[arXiv:1009.0657].
- [13] T. Sterling and J. Greensite Nucl. Phys. **B220** (1983) 327.
- [14] V. Azcoiti, E. Follana, A. Vaquero and G. Di Carlo JHEP 08 (2009) 008 [arXiv:0905.0639].
- [15] M. Göpfert and G. Mack Commun. Math. Phys. 82 (1981) 545.
- [16] M. Loan, M. Brunner, C. Sloggett and C. Hamer *Phys. Rev.* **D68** (2003) 034504 [arXiv:hep-lat/0209159].