

# Simulating the Random Surface Representation of Abelian Gauge Theories

Tomasz Korzec, Ulli Wolff



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MAINZ, GERMANY

based on [T.K., Ulli Wolff, Nucl.Phys. B871 (2013) 145]

# Ising Gauge Theory

## First and simplest lattice gauge theory

[Wegner, J.Math.Phys. 12 (1971) 2259]

- $D$ -dimensional hypercubic lattice of size  $L_0 \times L_1 \times \dots \times L_{D-1}$
- Gauge field  $\sigma(x, \mu) = \pm 1$  defined on the links
- Plaquette action

$$S[\sigma] = -\beta \sum_{x, \mu < \nu} \sigma(x, \mu) \sigma(x + \hat{\mu}, \nu) \sigma(x + \hat{\nu}, \mu) \sigma(x, \nu)$$

- Observables

$$\begin{aligned} \langle O[\sigma] \rangle &= \frac{1}{Z} \sum_{\{\sigma\}} O[\sigma] e^{-S[\sigma]} \\ \langle 1 \rangle &= 1 \quad \Rightarrow Z \end{aligned}$$

A general observable  $O[j] = \prod_{x, \mu} \sigma(x, \mu)^{j(x, \mu)}$  with  $j(x, \mu) \in \{0, 1\}$ .

Non-vanishing: closed loops or pairs of Polyakov lines

# Strong Coupling Expansion

$$Z[j] = \sum_{\{\sigma\}} e^{-S[\sigma]} \prod_{x,\mu} \sigma(x,\mu)^{j(x,\mu)}$$

Observables:  $\langle O[j] \rangle = Z[j]/Z[0]$

On every plaquette use:

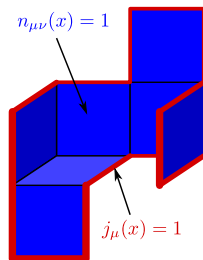
$$\begin{aligned} e^{\beta\sigma_p} &= \cosh(\beta) + \sigma_p \sinh(\beta) \\ &= \cosh(\beta) \sum_{n \in \{0,1\}} t^n \sigma_p^n, \quad t = \tanh(\beta) \end{aligned}$$

$$\begin{aligned} Z[j] &\propto \sum_{\{n\}} \left[ \prod_{x,\mu < \nu} t^{n_{\mu\nu}(x)} \right] \prod_{x,\mu} \left[ \sum_{\sigma} \sigma^{j(x,\mu) + \sum_{\nu > \mu} (n_{\mu\nu}(x) + n_{\mu\nu}(x-\hat{\nu}))} \right] \\ &\propto \sum_{\{n\}} \delta(\partial n + j) t^{\sum_{\mu < \nu} n_{\mu\nu}(x)} \end{aligned}$$

# Strong Coupling Expansion

The constraint  $\delta(\partial n + j)$ :

- Links with  $j = 0$  touch an even number of  $n = 1$  plaquettes
- Links with  $j = 1$  touch an odd number of  $n = 1$  plaquettes

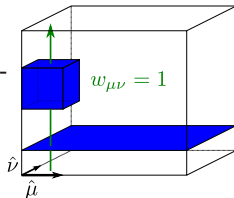


Topological sectors defined by “wrapping numbers”

$$w_{\mu\nu} = \sum_{x|_{x_{\mu}=x_{\nu}=0}} n_{\mu\nu}(x) \pmod{2}$$

Twisted boundary conditions can be incorporated by inclusion of the phase  $\Phi_{\gamma}$  into the graph weight

$$\Phi_{\gamma}[n] = \prod_{\mu < \nu} (\gamma_{\mu\nu})^{w_{\mu\nu}[n]}$$



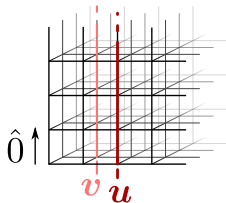
where  $\gamma_{\mu\nu}$  is the twist in the  $\mu/\nu$  plane



# Enlarged Ensemble

- Polyakov line at spatial position  $\mathbf{u}$ :  
$$\pi(\mathbf{u}) = \prod_{u_0} \sigma(\mathbf{u}, 0)$$
- Defects  $j^{(\mathbf{u}, \mathbf{v})}$ , such that  $O[j^{(\mathbf{u}, \mathbf{v})}] = \pi(\mathbf{u})\pi(\mathbf{v})$
- Simulate enlarged ensemble

$$\begin{aligned} \mathcal{Z} &= \sum_{\mathbf{u}, \mathbf{v}} \rho^{-1}(\mathbf{u} - \mathbf{v}) Z[j^{(\mathbf{u}, \mathbf{v})}] \\ &= \sum_{\{n\}, \mathbf{u}, \mathbf{v}} \rho^{-1}(\mathbf{u} - \mathbf{v}) t^{\sum_{x, \mu, \nu} n_{\mu\nu}(x)} \delta(\partial n + j^{(\mathbf{u}, \mathbf{v})}) \end{aligned}$$



- Observables

$$\langle\langle O[n, \mathbf{u}, \mathbf{v}] \rangle\rangle = \frac{1}{\mathcal{Z}} \sum_{\{n\}, \mathbf{u}, \mathbf{v}} O[n, \mathbf{u}, \mathbf{v}] \rho^{-1}(\mathbf{u} - \mathbf{v}) t^{\sum_{x, \mu, \nu} n_{\mu\nu}(x)} \delta(\partial n + j^{(\mathbf{u}, \mathbf{v})})$$

# Observables

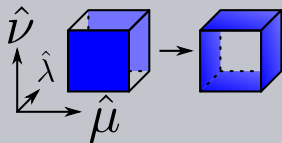
- Polyakov correlator:  $G(\mathbf{x}) = \langle \pi(\mathbf{x})\pi(\mathbf{0}) \rangle = \rho(\mathbf{x}) \frac{\langle \langle \delta_{\mathbf{x}, \mathbf{u}-\mathbf{v}} \rangle \rangle}{\langle \langle \delta_{\mathbf{u}, \mathbf{v}} \rangle \rangle}$   
→ simple counters!
- Abbreviation for “vacuum observables”  $\langle \langle O[n] \rangle \rangle_0 \equiv \frac{\langle \langle O[n] \delta_{\mathbf{u}, \mathbf{v}} \rangle \rangle}{\langle \langle \delta_{\mathbf{u}, \mathbf{v}} \rangle \rangle}$
- Average plaquette  $\langle \sigma_p \rangle = \frac{\partial \log Z}{\partial \beta} = t + \frac{(t^{-1} - t)}{N_p} \left\langle \left\langle \sum_{\mathbf{x}, \mu < \nu} n_{\mu\nu}(\mathbf{x}) \right\rangle \right\rangle_0$   
→ Related to the total surface of vacuum graphs
- Topological observables  $\frac{Z_\gamma}{Z} = \langle \langle \prod_{\mu < \nu} (\gamma_{\mu\nu})^{w_{\mu\nu}[n]} \rangle \rangle_0$ 
  - ▶ Do not need renormalization
  - ▶ Can be obtained from the same simulation
  - ▶ Very sensitive to  $\beta$
  - ▶ Higher derivatives  $\partial/\partial\beta$  also feasible

# Simulation: Cube Updates

## A defect-conserving update ( $D > 2$ )

- Proposal

- ▶ Choose a 3D cube, i.e.  $x$  and  $\mu < \nu < \lambda$  (randomly or in an ordered fashion)
- ▶ The six plaquettes associated with this cube are  $\mathcal{P} = \{(x, \mu, \nu), (x + \hat{\lambda}, \mu, \nu), (x, \mu, \lambda), (x + \hat{\nu}, \mu, \lambda), (x, \nu, \lambda), (x + \hat{\mu}, \nu, \lambda)\}$
- ▶ propose to flip these six plaquettes:  
 $n(p) \rightarrow 1 - n(p)$  for all  $p \in \mathcal{P}$



- Acceptance with Metropolis probability  $\min[1, q]$

$$q = \prod_{p \in \mathcal{P}} t^{1-2n(p)}$$

- Ancient! [T.Sterling, J.Greensite, Nucl.Phys.B220 (1983)]
- Does not change  $\partial n$ . I.e is not ergodic
- Does not change the wrapping number  $w_{\mu\nu}$   
→ Fluctuating boundary conditions only
- In  $D = 3$ : related to a spin-flip in the dual Ising model  
→ critical slowing down with  $z \approx 2$  expected.

# Simulation: Cube-Cluster Update

Instead of sweeps of cube-flips one can build clusters of cubes

## Cluster Algorithm

- Determine for every plaquette a bond  $b(p) \in \{0, 1\}$ . The value 1 has probability
$$P(b(p) = 1) = 1 - t^{1-n(p)}$$
  - Choose a 3D cube  $c_0$  randomly:  $x, \mu < \nu < \lambda$
  - Determine the maximal set  $C$  of cubes, so that for each cube  $c \in C$ 
    - ▶ There exists a path from  $c$  to  $c_0$  that crosses only plaquettes with  $b(p) = 1$
    - ▶ The entire path lies in the 3D subspace spanned by  $\hat{\mu}, \hat{\nu}, \hat{\lambda}$
  - For every  $c \in C$ , flip  $n(p) \rightarrow 1 - n(p)$  on all six plaquettes  
→ Amounts to flipping  $n$  on the cluster boundary
- 
- Eliminates critical slowing down in  $D = 3$  with fluctuating b.c.
  - Is a valid algorithm in  $D > 3$ , but is it efficient?

# Simulation: Polyakov Shift Updates

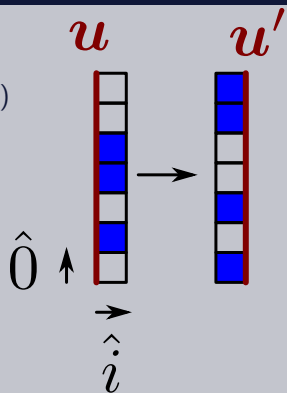
Move the Polyakov lines while preserving the constraints

## Worm Algorithm

- Calculate auxiliary field

$$k(\mathbf{x}, i) = \sum_{x_0=0}^{L_0-1} n_{0i}((x_0, \mathbf{x})), \quad k(\mathbf{x}, -i) \equiv k(\mathbf{x} - \hat{i}, i)$$

- If  $\mathbf{u} = \mathbf{v}$ , move both to a random position
- Propose to move  $\mathbf{u}$  to one of its  $2(D-1)$  spatial neighbors  $\mathbf{u} \rightarrow \mathbf{u}' = \mathbf{u} \pm \hat{i}$ , together with a flip of all  $n$  in the “ladder”
- Accept proposal with probability  $\min \left[ 1, t^{L_0 - 2k(\mathbf{u}, \pm i)} \frac{\rho(\mathbf{u} - \mathbf{v})}{\rho(\mathbf{u}' - \mathbf{v})} \right]$



- Together with cube-flips: ergodic
- Can change the wrapping numbers  $w_{0i}$

# Simulation: Non-Rejecting Polyakov Shift Updates

- Potential problem: High rejection rate in worm updates when  $L_0$  is large
- Can be cured, if we are only interested in vacuum observables  $\langle\langle \mathcal{O} \rangle\rangle_0$

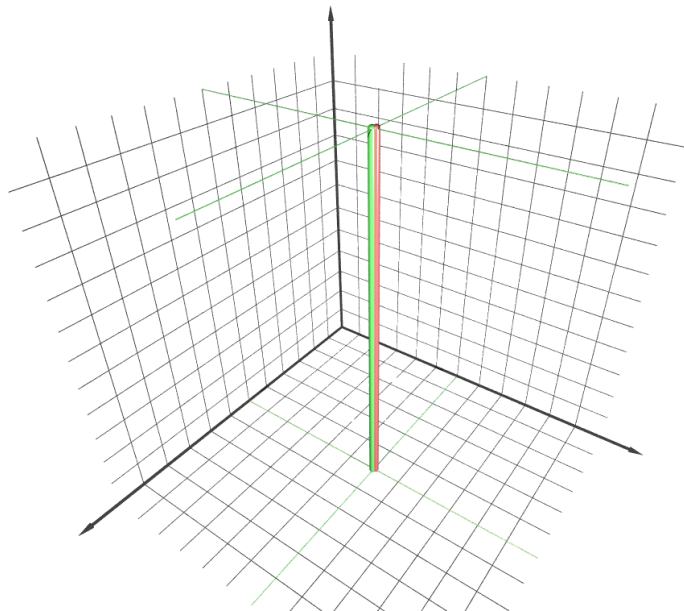
[Q.Liu, Y.Deng, T.Garoni, Nucl.Phys. B846 (2011)]

## Rejection Free Update

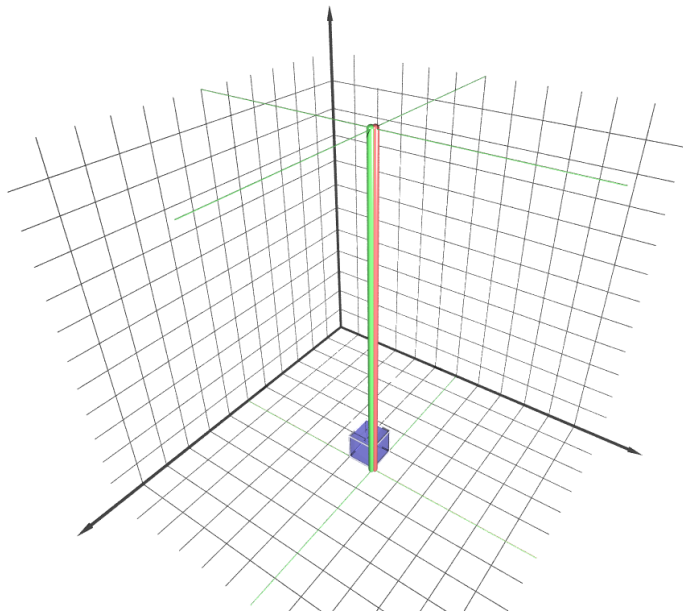
- If  $\mathbf{u} = \mathbf{v}$ : move  $\mathbf{u}$  according to the usual worm algorithm
- If  $\mathbf{u} \neq \mathbf{v}$ :
  - ▶ Calculate usual worm acceptance probabilities for all directions  $i$   
$$\tilde{p}_{i,\pm} = \min \left[ 1, t^{L_0 - 2k(\mathbf{u}, \pm i)} \frac{\rho(\mathbf{u} - \mathbf{v})}{\rho(\mathbf{u}' - \mathbf{v})} \right]$$
  - ▶ Calculate  $A = \frac{1}{2(D-1)} \sum_i (\tilde{p}_{i,+} + \tilde{p}_{i,-})$
  - ▶ Move  $\mathbf{u}$  to position  $\mathbf{u} + \hat{i}$  with probability  $p_{i,\pm} = \frac{\tilde{p}_{i,\pm}}{2(D-1)A}$   
 $\sum p_{i,\pm} = 1 \rightarrow$  a move is always made
- The sampled ensemble changes to

$$\mathcal{Z}_{\text{nr}} = \sum_{\{n\}, \mathbf{u}, \mathbf{v}} t^{\sum_p n(p)} \rho^{-1}(\mathbf{u} - \mathbf{v}) \delta(\partial n + j^{(\mathbf{u}, \mathbf{v})}) [\delta_{\mathbf{u}, \mathbf{v}} + (1 - \delta_{\mathbf{u}, \mathbf{v}})A]$$

# Ising Gauge Theory Simulation Example

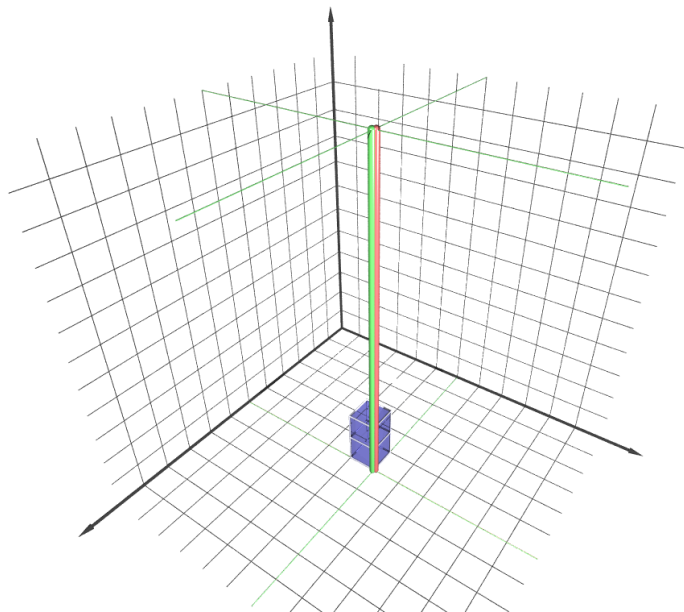


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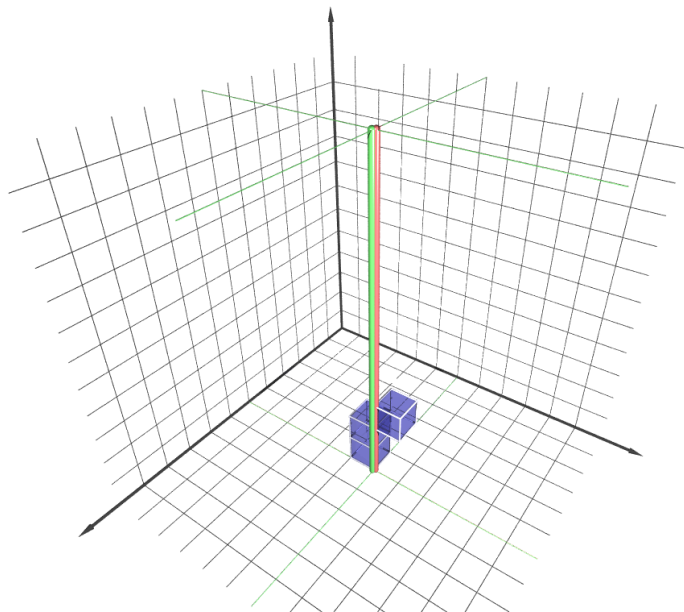




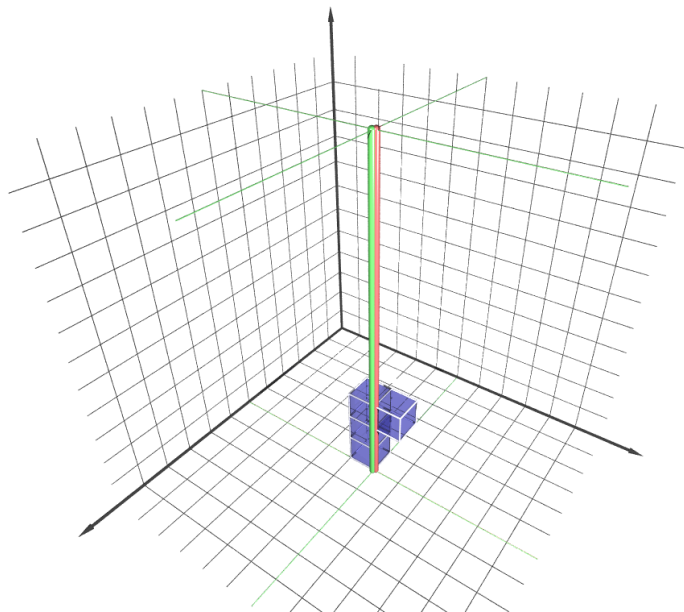
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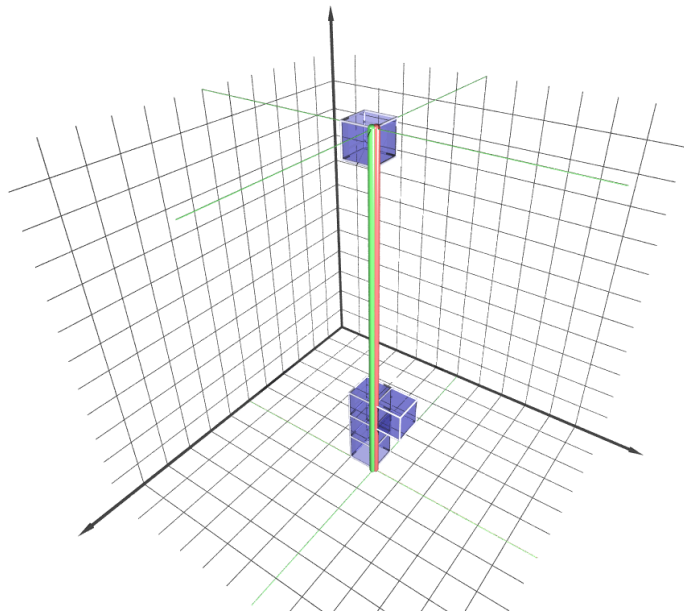
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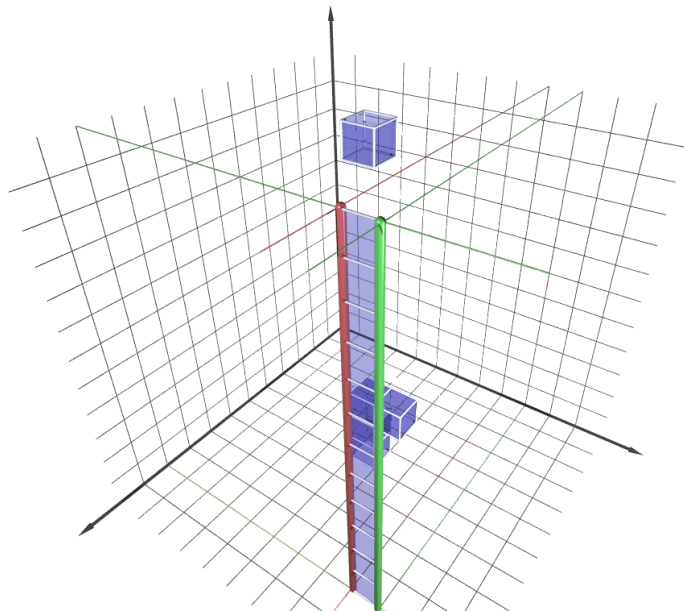
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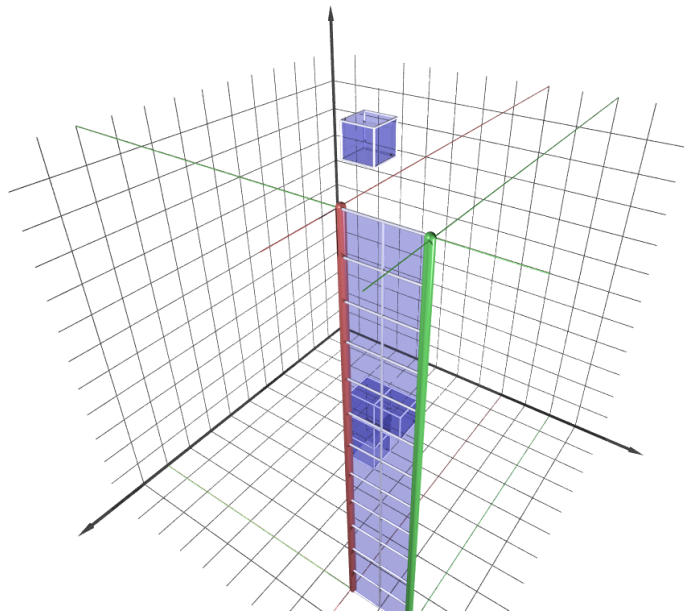
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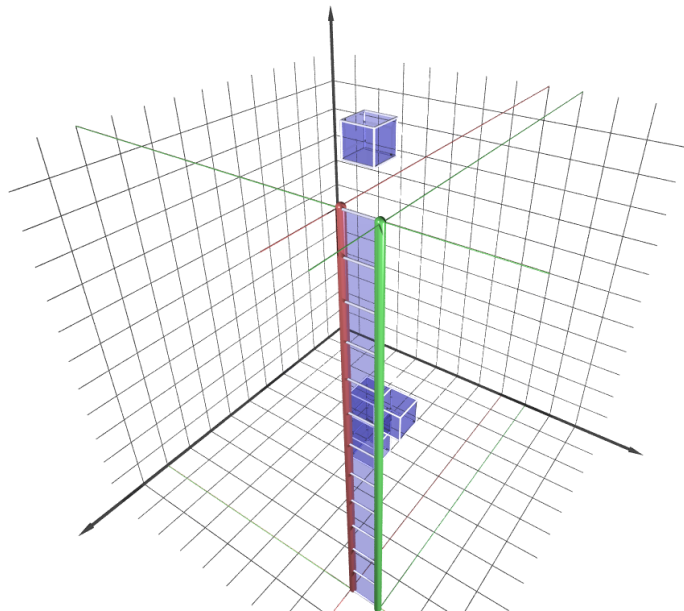
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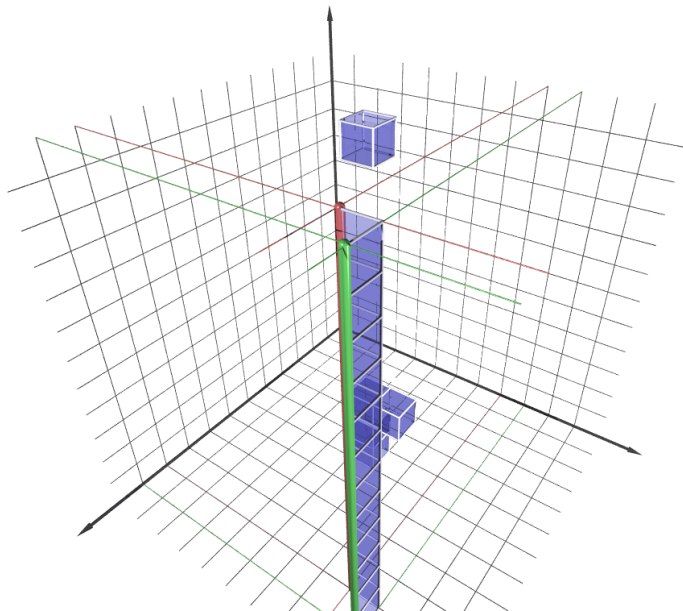
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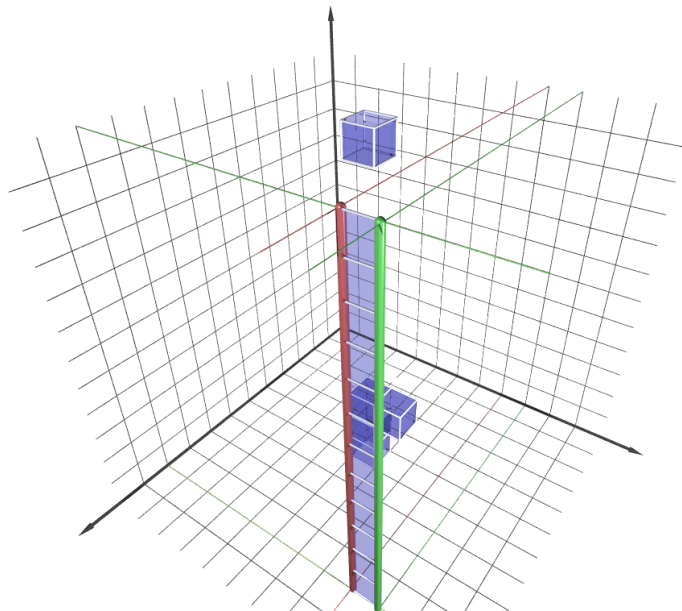


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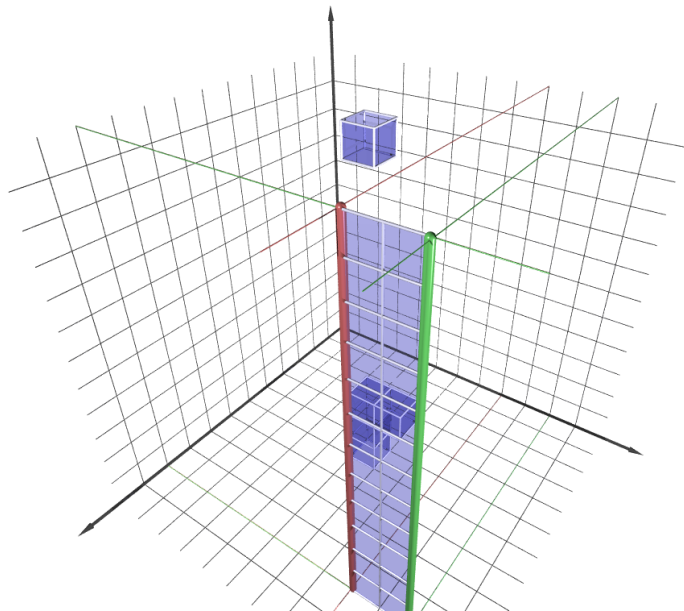




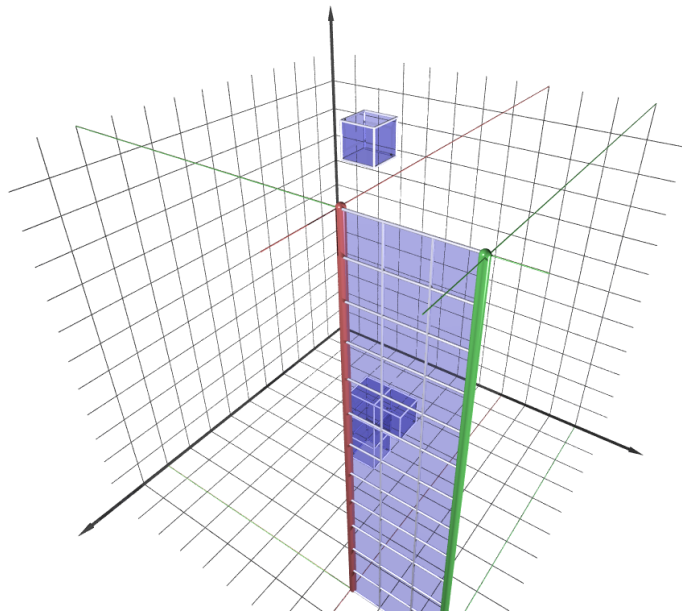
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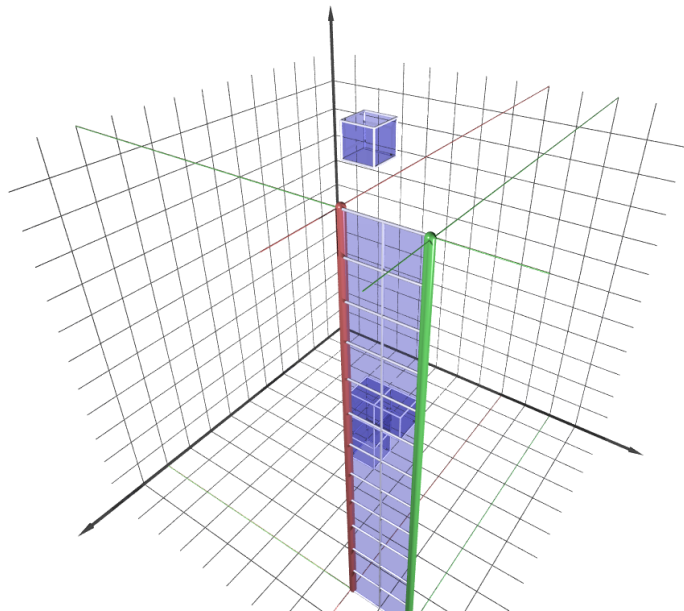
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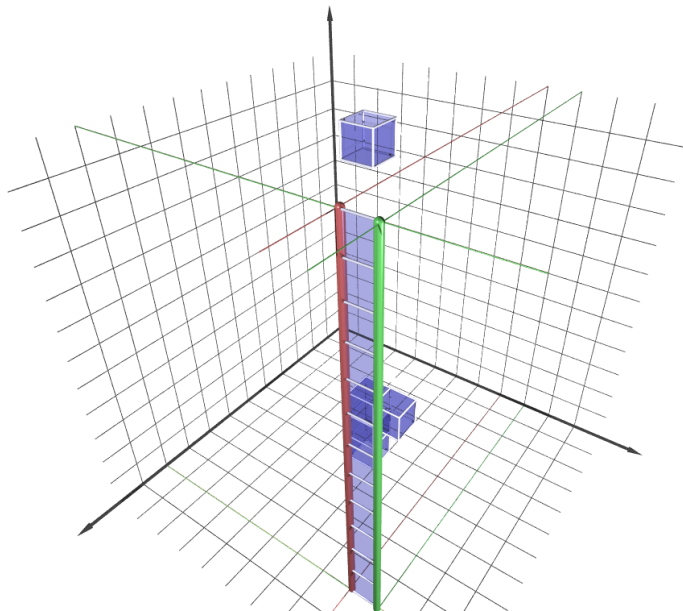
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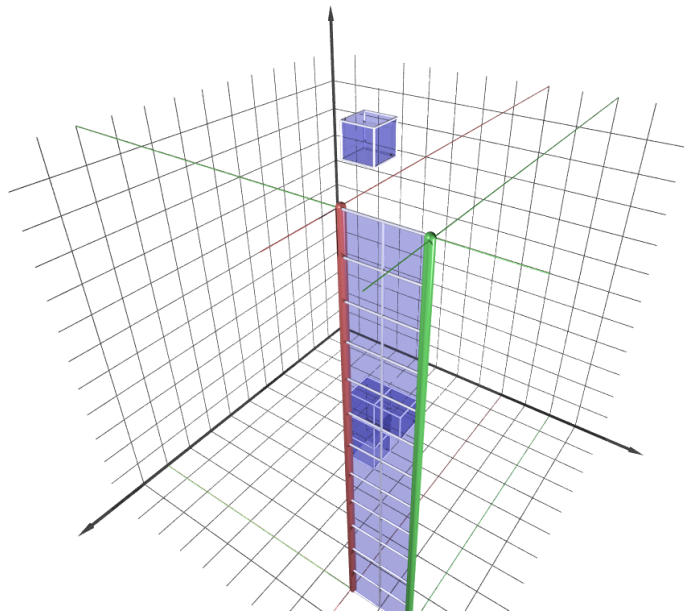
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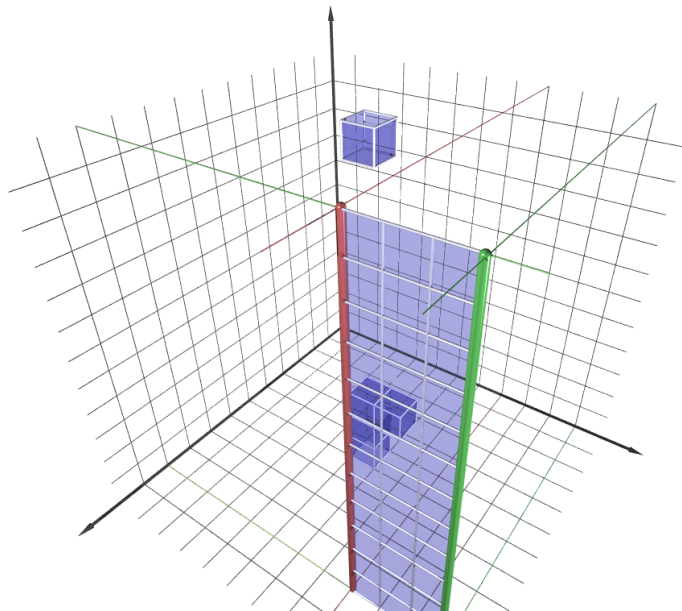
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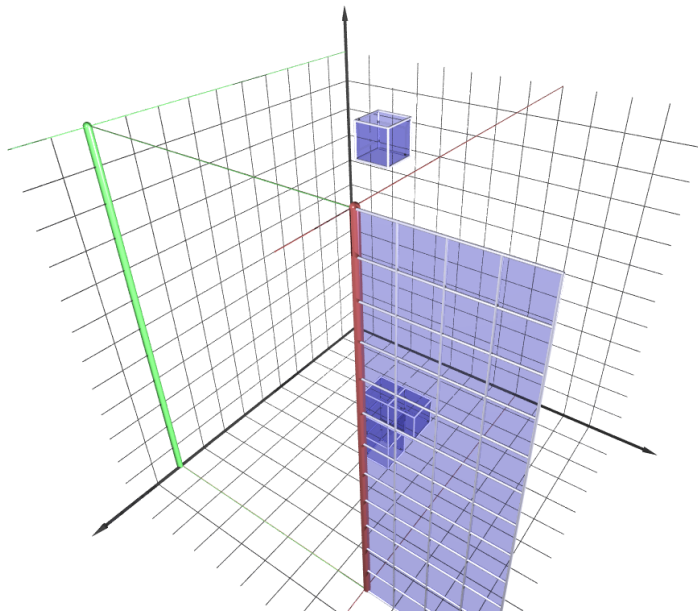
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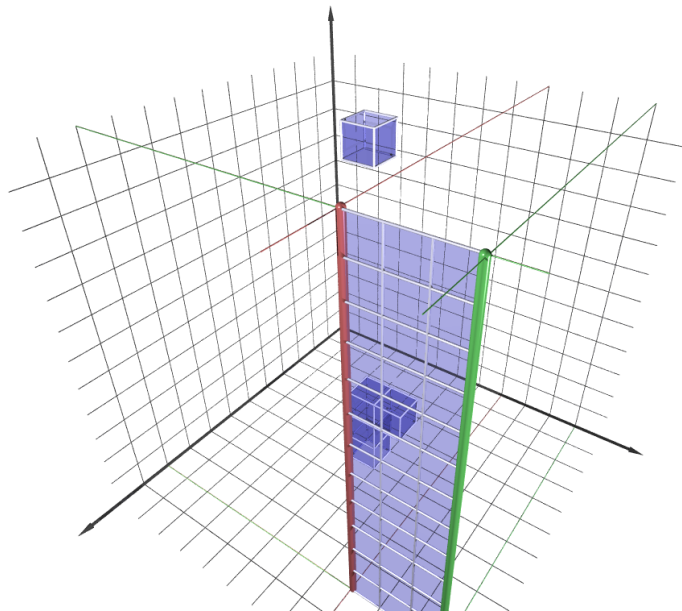


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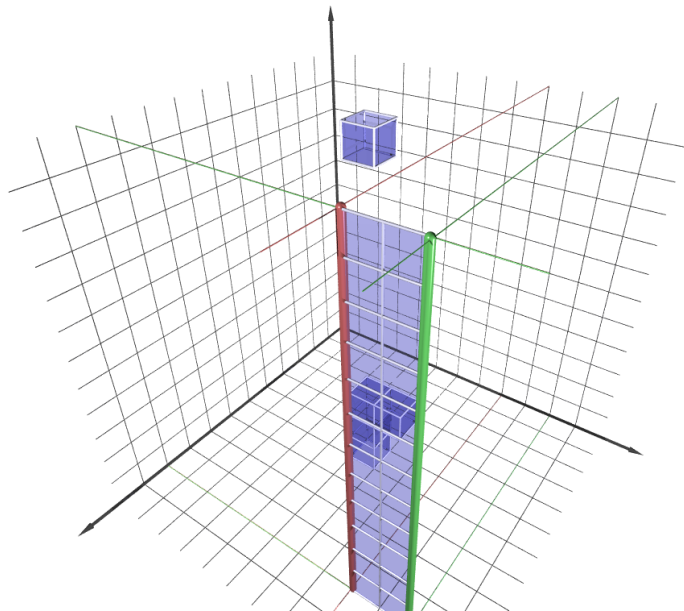




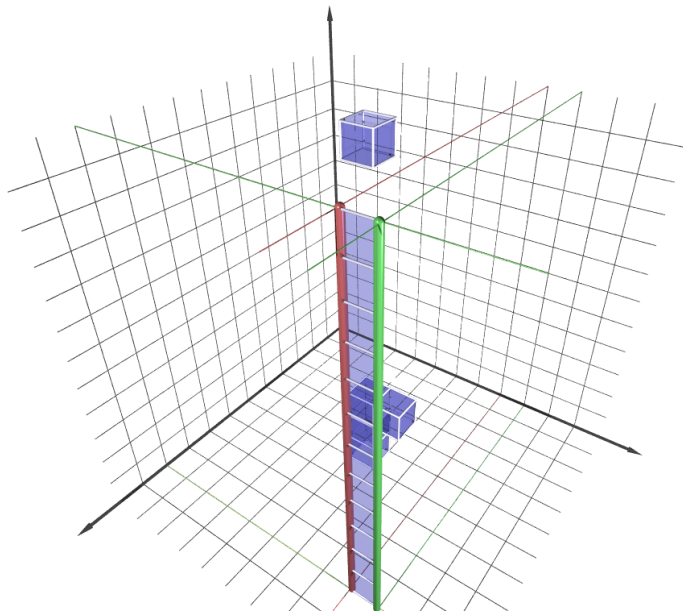
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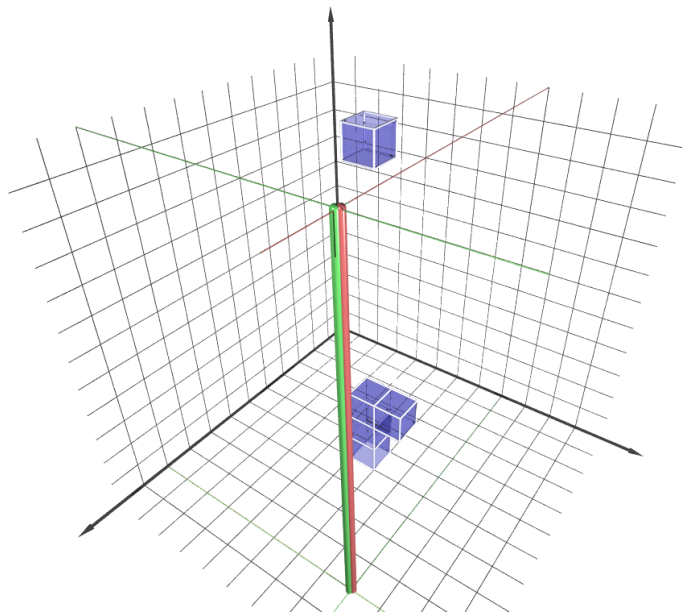
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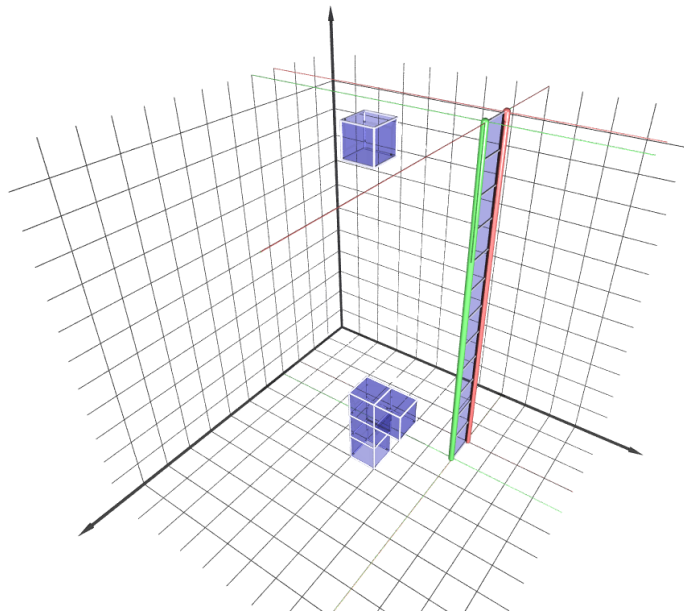
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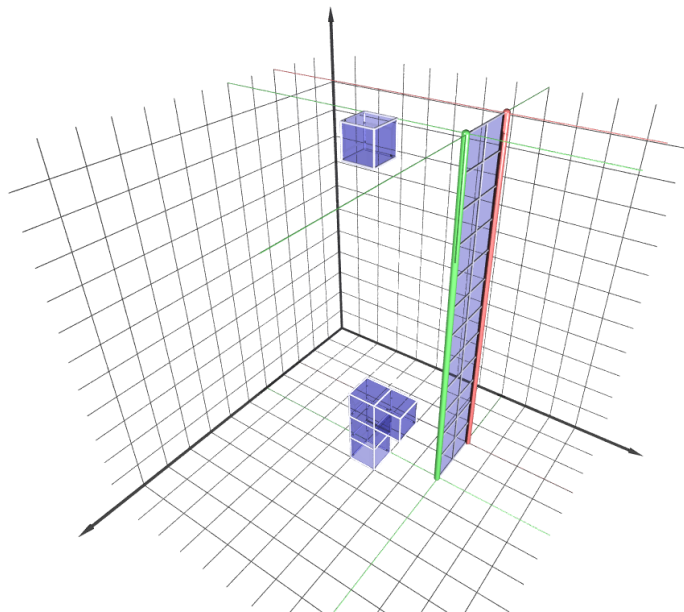
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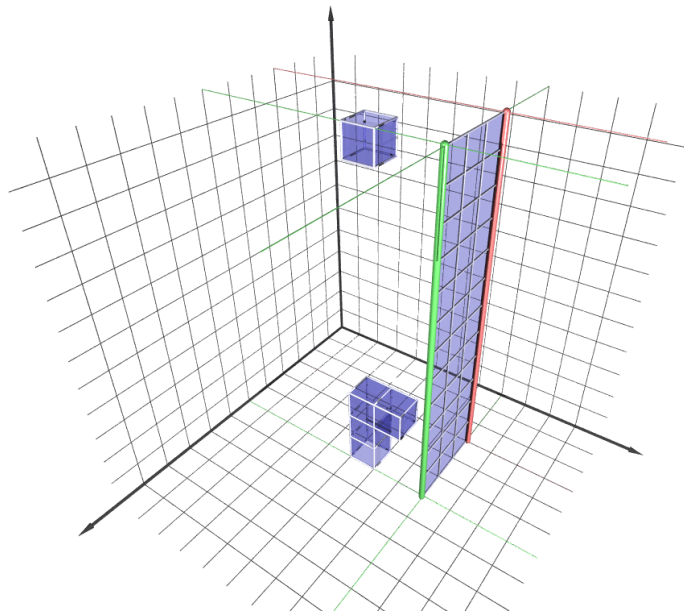
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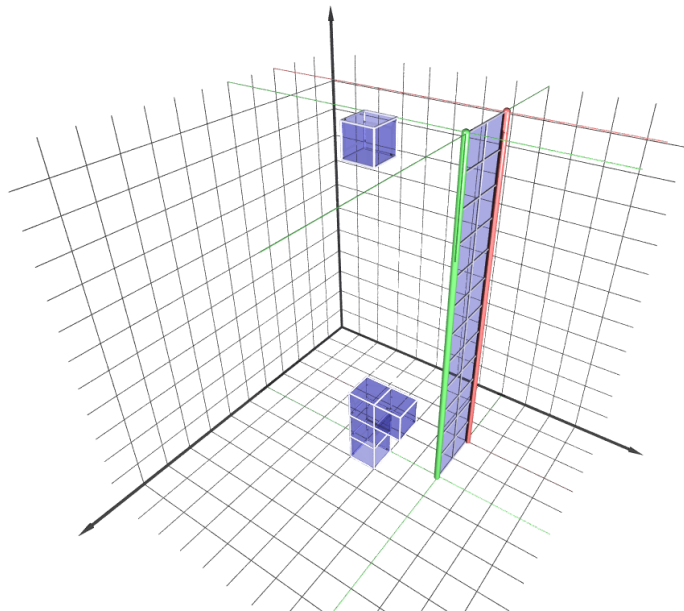
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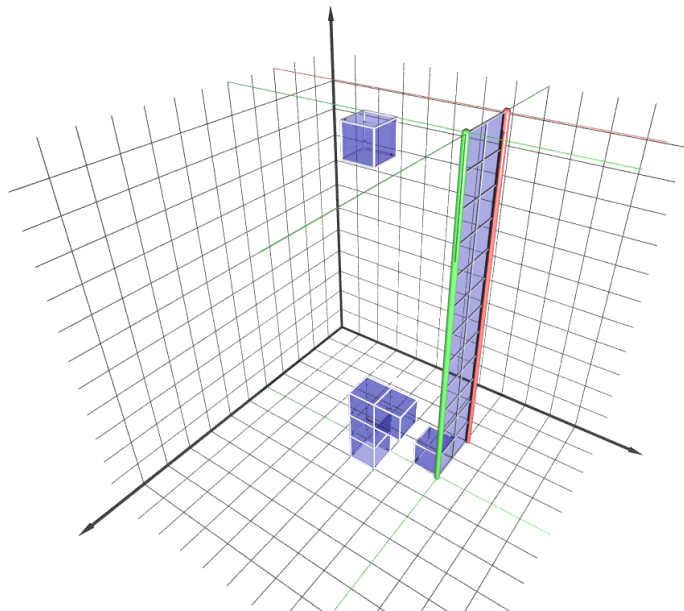


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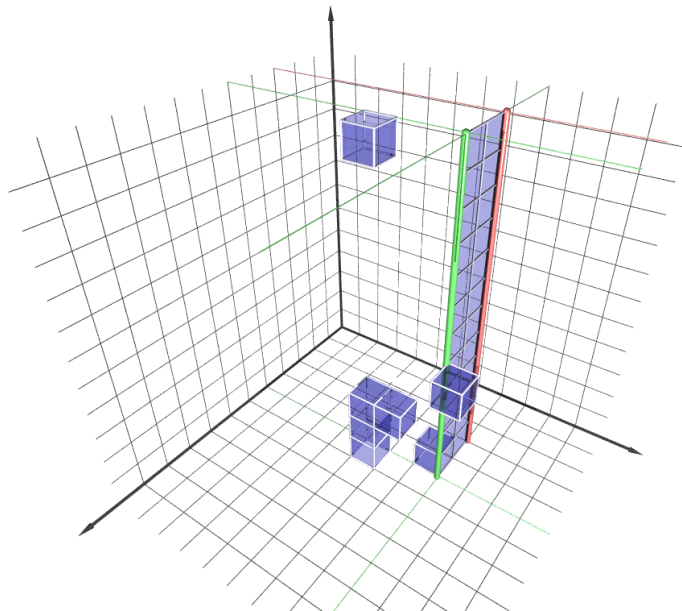




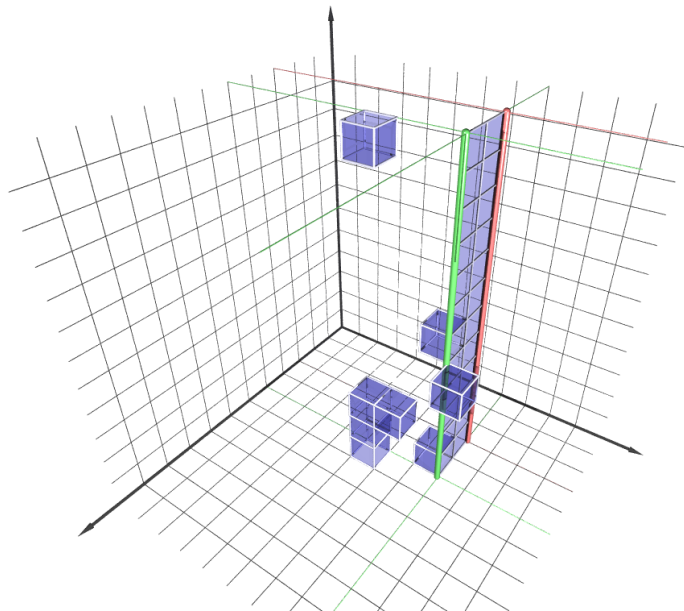
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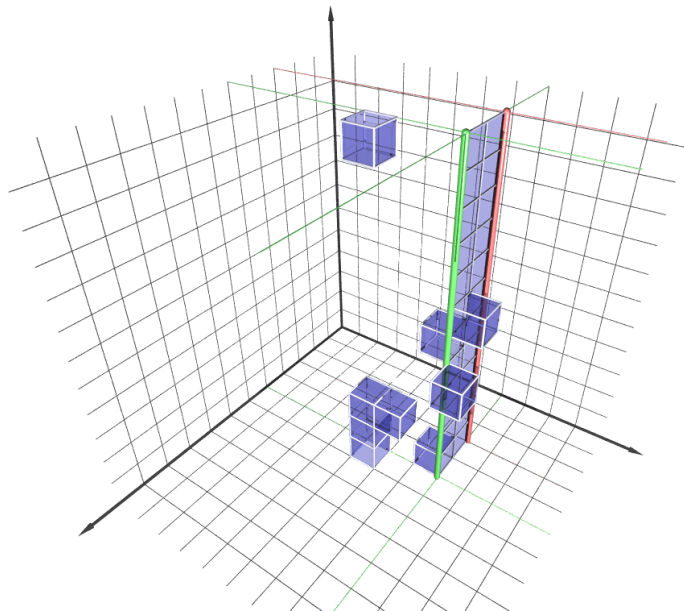
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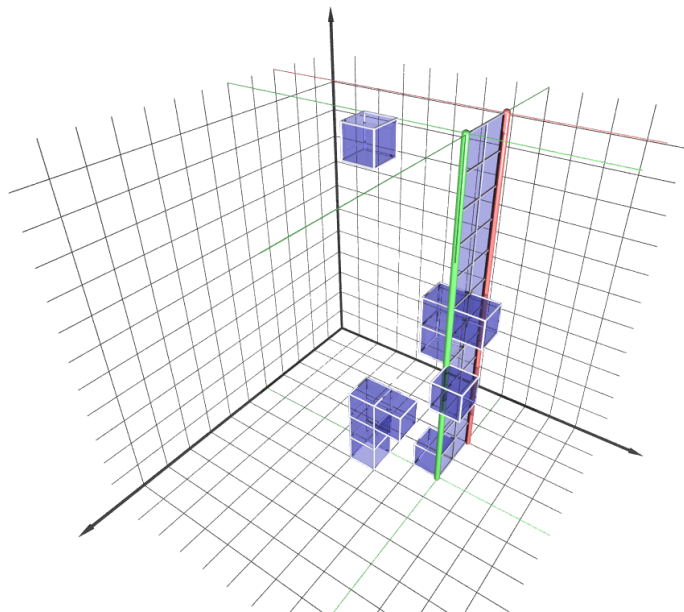
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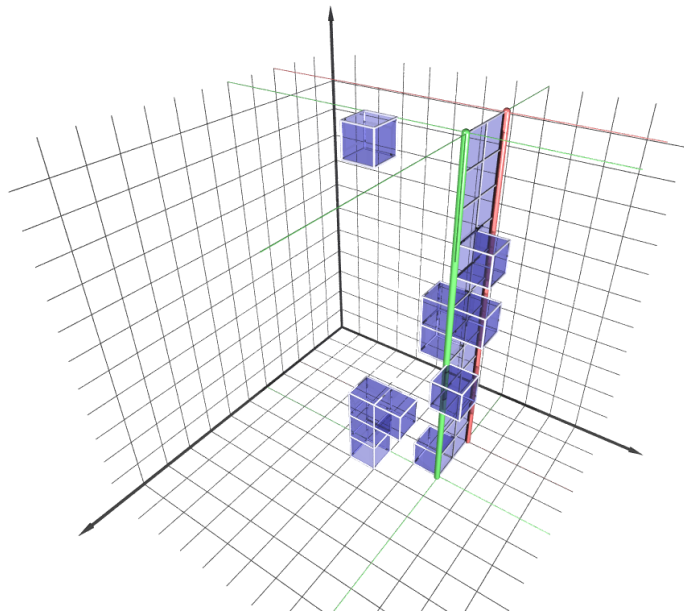
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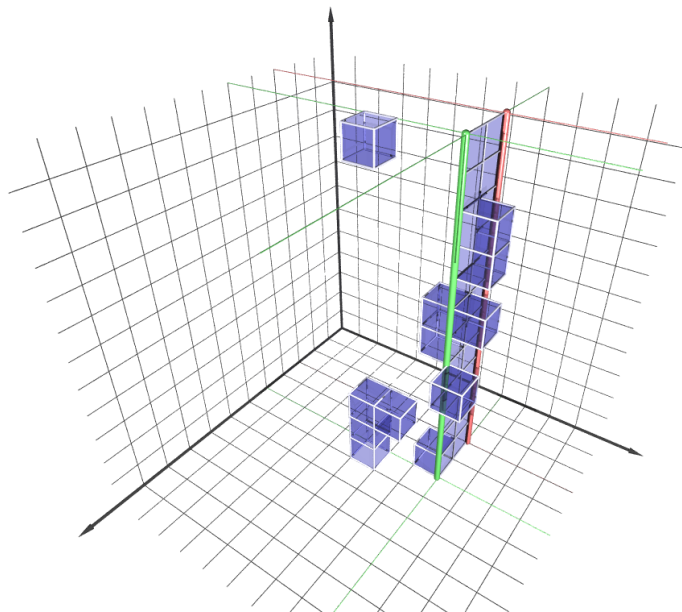
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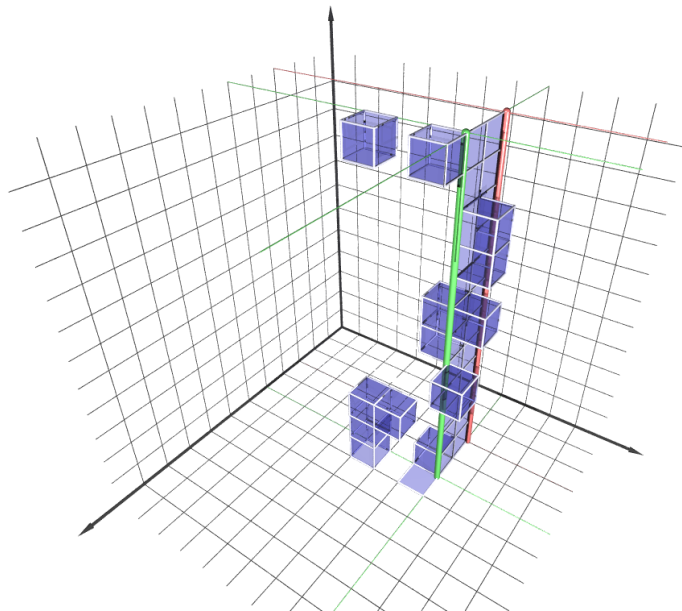
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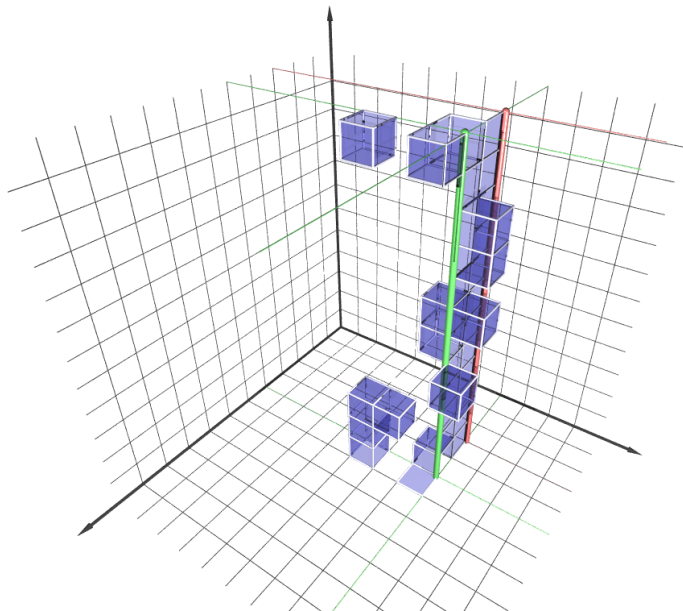


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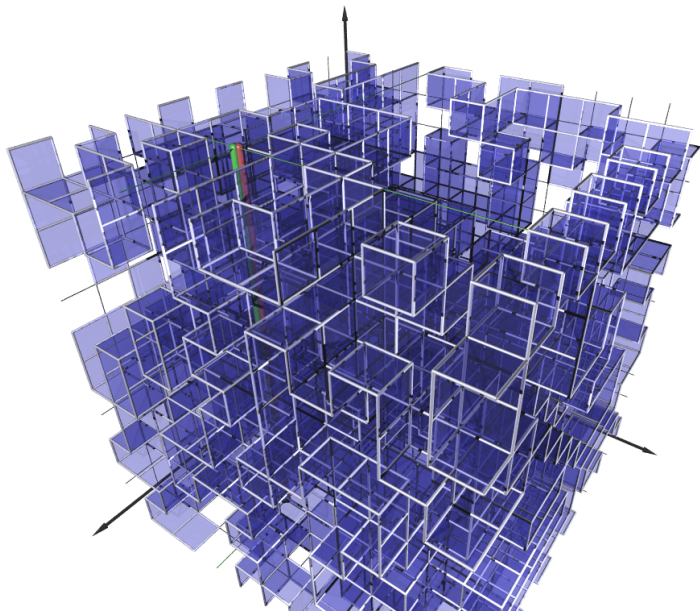




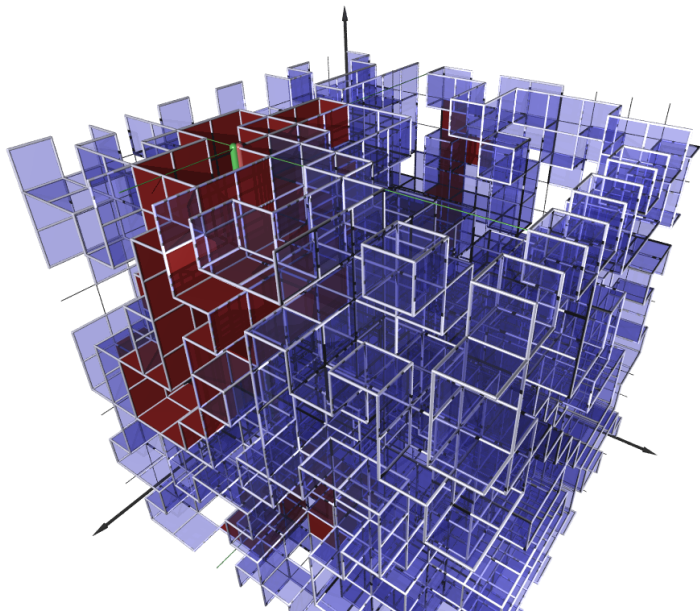
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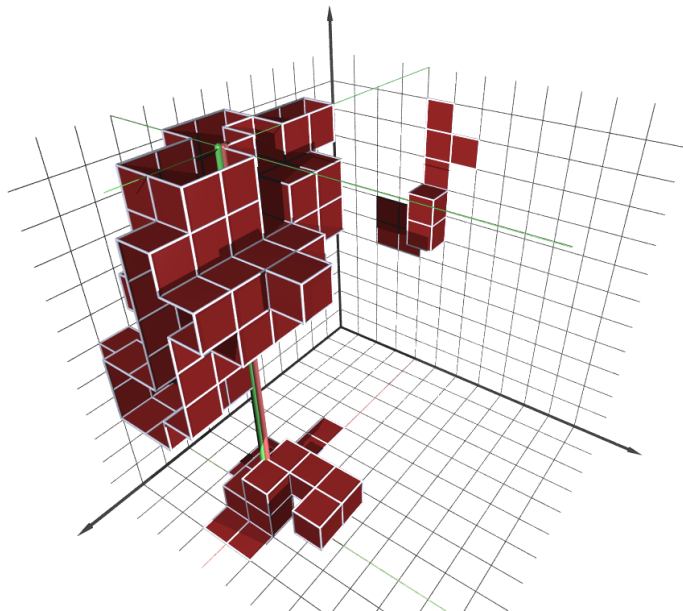
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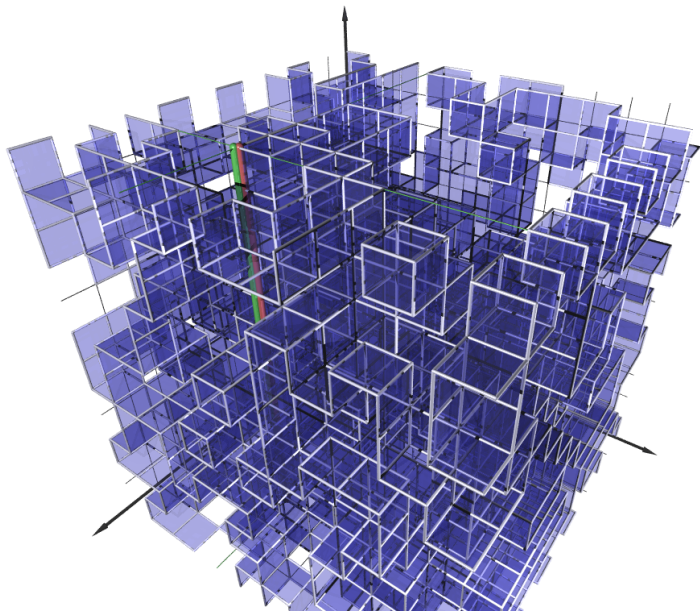
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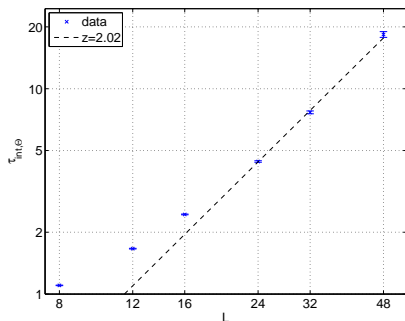


# Critical Slowing Down

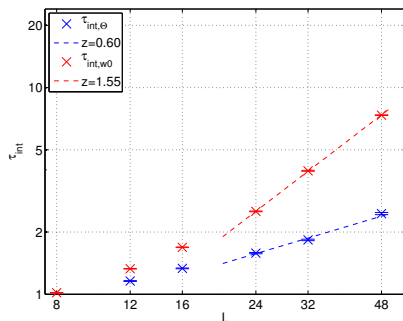
$\beta = 0.76141346 \approx \beta_c$  known from the dual Ising model

[M.Hasenbusch, Phys.Rev.B85 174421 (2012)], [Y.Deng, H.Blöte, Phys.Rev. E68 036125 (2003)]

## Cubes + NR-Shifts



## Cubes + Cluster + NR-Shifts



- At critical point:  $\tau_{\text{int}} \sim L^z$
- $\Theta$  is the total surface in  $u = v$  configurations.  
 $w_0$  is the fraction of zero-wrapping number configurations

# Simulations in the Confined Phase

- $\beta < \beta_c$ , anisotropic lattice  $L_0 \times L \times L$
- We measure the correlator  $G(\mathbf{x}) = \langle \pi(\mathbf{x})\pi(\mathbf{0}) \rangle$
- Project to zero momentum:  $C(x_1) = \sum_{x_2} G(\mathbf{x})$
- $C(x_1) = \sum_n |v_n|^2 \exp(-E_n x_1)$
- We are mainly interested in  $E_0$  as a function of  $L_0$   
 $\Rightarrow$  effective string theory
- Infinite volume, continuum:

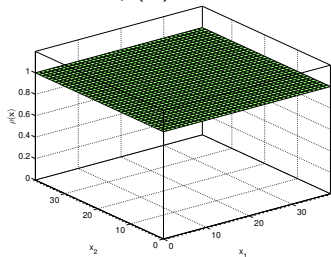
[M.Lüscher, P.Weisz, JHEP 0407, 014 (2004)]

$$G(\mathbf{x}) = \sum_{n \geq 0} |v_n|^2 2r \left( \frac{E_n}{2\pi r} \right)^{\frac{D-1}{2}} K_{\frac{1}{2}(D-3)}(E_n r)$$

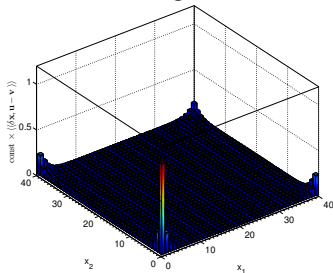
- Inspired by this formula we set the weight function to  $\rho(\mathbf{x}) \propto \sum_{\mathbf{k} \in \mathbb{Z}} K_0 \left( \hat{M} \sqrt{(x_1 - k_1 L)^2 + (x_2 - k_2 L)^2} \right)$   
with one free algorithmic parameter  $\hat{M}$

# Polyakov Correlators: Reweighting

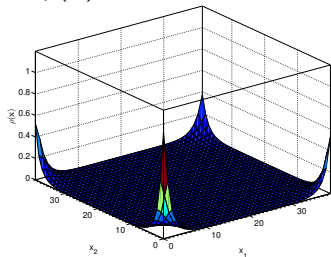
$$\rho(\mathbf{x}) = 1$$



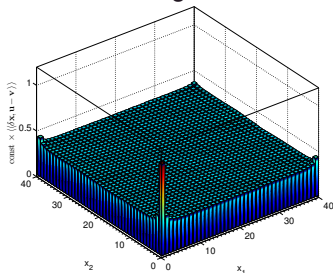
Histogram



$$\rho(\mathbf{x}) \text{ with } \hat{M} = 0.4825$$



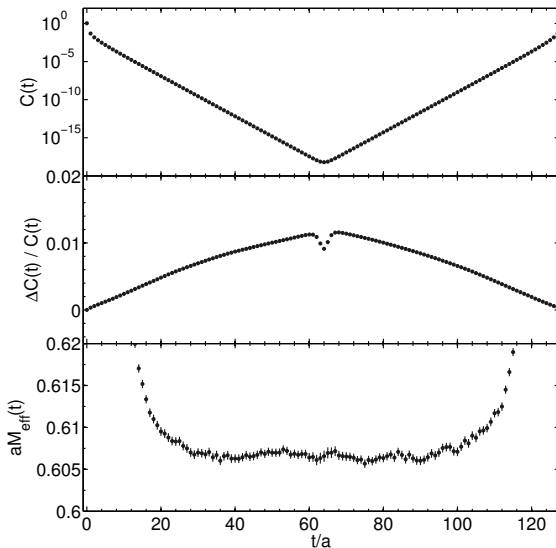
Histogram





# Closed String Mass Gap

- $L_0 = 56, L = 128$
- $M_{\text{eff}}(t) = \text{acosh} \left[ \frac{C(t-1) + C(t+1)}{2C(t)} \right]$
- $E_0 = 0.60661(24)$



# Conclusions and Outlook

- Algorithm

- ▶ Random surface formulation of Ising gauge theory with positive, factorizable weights
- ▶ Simulation with local and cluster algorithms
- ▶ Straightforward adjustment to e.g.  $U(1)$  gauge theory
- ▶ Critical slowing down not eliminated
- ▶ But: other beneficial worm properties salvaged
  - ★ Interesting observables that are hard to measure in a traditional simulation
  - ★ Modified importance sampling to interesting observables

- Physics

- ▶ Effective string theory picture of confinement → Ulli Wolff's talk