

## Random Walks

A random walk is a tuple of  $T$   $d$ -dimensional steps.

- Random Walk (RW): Single coordinates of steps are drawn from a Gaussian distribution  $G(0, 1)$ .
- Lattice Random Walk (LRW): Every step moves to an adjacent lattice site. It may only traverse one site per step. Here, the lattice is a square lattice. Immediate reversals are allowed.
- Self-Avoiding Walk (SAW): A LRW but steps may not visit already visited sites.
- Loop-Erased Random Walk (LERW): A LRW but loops are erased from the walk such that there are no intersections.

One-dimensional properties scale like  $T^\nu$ .

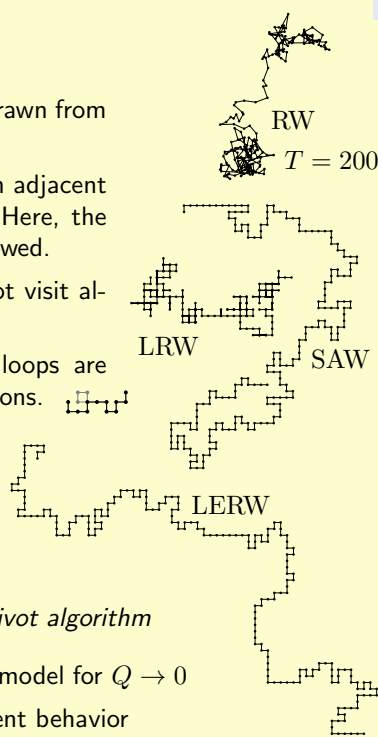
- $1/2 = \nu_{\text{RW}} \leq \nu_{\text{SAW}} \leq \nu_{\text{LERW}}$ , equality for  $d \geq 4$ .

SAW: Model for polymers [1]

- not trivial to generate uniformly distributed instances
- use Markov chain to generate new realizations, i.e., *pivot algorithm*

LERW: Connected to uniform spanning trees and the Potts model for  $Q \rightarrow 0$

- intended as an easy version of SAW, but shows different behavior

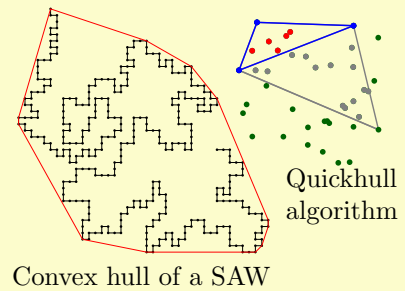


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## Convex Hulls

The convex hull of a set of points is the smallest convex polygon enclosing all  $T$  points.

- area  $A$  and perimeter  $L$  characterize the point set
- well researched in the context of computer graphics
- fast:  $\sim \mathcal{O}(T \log T)$
- easy generalization to points in  $d$  dimensions



Convex hull of a SAW

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## Large Deviation Simulation

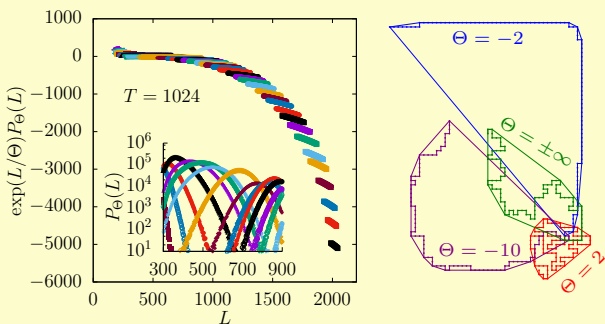
Using sophisticated sampling methods [2], the large deviation tails of the distributions, here with probabilities below  $P(L) = 10^{-800}$ , can be obtained.

- Markov chain Monte Carlo sampling using a "Temperature"  $\Theta$  based Metropolis algorithm
- propose a change of the configuration  $c_t \rightarrow c_{t+1}$ , accept depending on "energy" difference  $\Delta S$  (here:  $S$  is either area  $A$  or perimeter  $L$ )

$$p_{\text{acc}} = \min \left[ 1, e^{-\Delta S / \Theta} \right]$$

- after equilibration: sample  $S$
- biased distributions  $P_\Theta(S)$  can be transformed into the true distribution

$$P(S) = e^{S/\Theta} Z(\Theta) P_\Theta(S)$$



- calculate  $Z(\Theta)$  by matching overlaps
- can be applied to a wide range of problems and enhancements like parallel tempering are applicable [3]
- alternative: Wang Landau sampling

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## Distributions of Area and Perimeter

Distribution for SAW and LERW over whole support

- decent system sizes up to  $N = 2048$
- no small  $L$  and  $A$  because of excluded volume effects

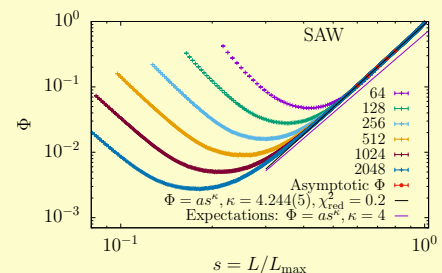
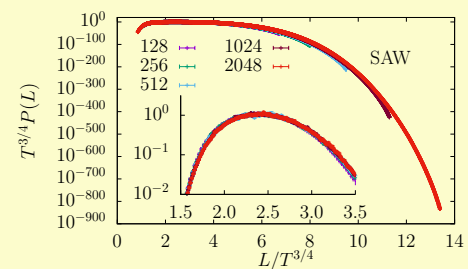
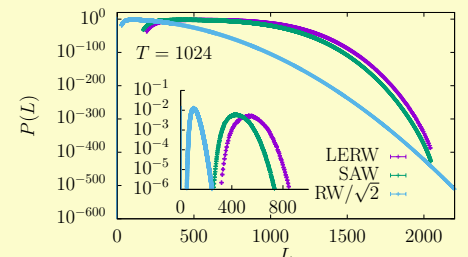
Whole distribution scales well with  $T^\nu$ , as in the RW case [4]

- deviations near  $S_{\text{max}}$  due to lattice structure
- excellent at the peak region

Empirical rate function

$$\Phi(S/S_{\text{max}}) = -\frac{1}{T} \log P(S)$$

- extrapolated from the large  $S$  tail for  $T \rightarrow \infty$  by a power law
- shows slight deviations from prediction  $\kappa = \frac{1}{d(1-\nu)}$ .
- fits better for  $L$  than for  $A$
- $A$  shows stronger effects of the lattice structure



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## Bibliography

- [1] N. Madras and G. Slade, "The Self-Avoiding Walk" (2013)
- [2] A.K. Hartmann, Phys. Rev. E **65**, 056102 (2002)
- [3] P. Fieth, A.K. Hartmann, Phys. Rev. E **94**, 022127 (2016)
- [4] G. Claussen, A.K. Hartmann, and S.N. Majumdar. Phys. Rev. E **91**, 052104 (2015)

