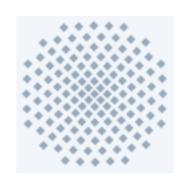
# Wang-Landau sampling for Quantum Monte Carlo

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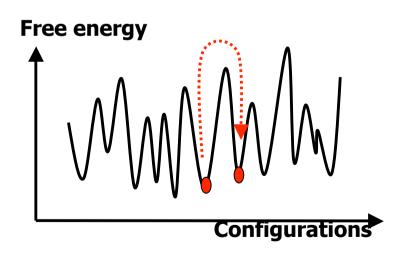
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## **Overview**

- Classical Monte Carlo
- First order phase transitions
- " Classical Wang-Landau algorithm
- Extension to quantum systems
- Applications



## **Classical Monte Carlo simulations**

We want to calculate a thermal average

$$\langle A \rangle = \sum_{c} A_{c} e^{-\beta E_{c}} / Z$$
 with  $Z = \sum_{c} e^{-\beta E_{c}}$ 

- Exponentially large number of configurations⇒ draw a representative statistical sample by importance sampling
  - Pick M configurations  $c_i$  with probability  $p_{c_i} = e^{-\beta E_{c_i}} / Z$
  - "Calculate statistical average  $\langle A \rangle \approx \overline{A} = \frac{1}{M} \sum_{i=1}^{M} A_{c_i}$
  - Within a statistical error  $\Delta A = \sqrt{\frac{\text{Var } A}{M}}$
- " Problem: we cannot calculate  $p_{c_i} = e^{-\beta E_{c_i}}/Z$  since we do not know Z

# Markov chains and Metropolis algorithm

" Metropolis algorithm builds a Markov chain

$$c_1 \rightarrow c_2 \rightarrow \dots \rightarrow c_i \rightarrow c_{i+1} \rightarrow \dots$$

- Transition probabilities  $W_{x,y}$  for transition  $x \rightarrow y$  need to fulfill
  - Ergodicity: any configuration reachable from any other

$$\forall x, y \; \exists n \; : \; \left(W^n\right)_{x,y} \neq 0$$

Detailed balance:

$$\frac{W_{x,y}}{W_{y,x}} = \frac{p_y}{p_x}$$

" Simplest algorithm due to Metropolis (1953):

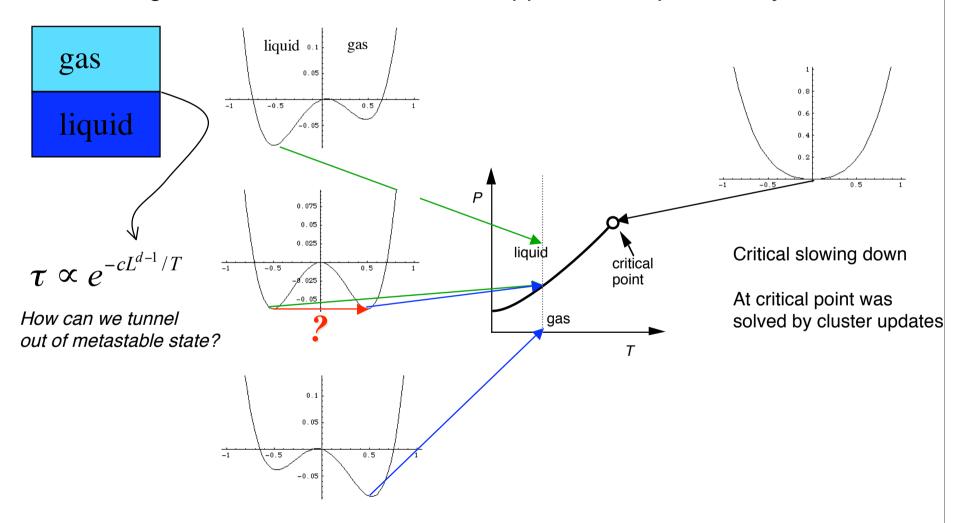
$$W_{x,y} = \min[1, p_y / p_x]$$

" Needs only relative probabilities (energy differences)

$$p_{y}/p_{x} = e^{-\beta(E_{y}-E_{x})}$$

# First order phase transitions

Tunneling out of meta-stable state is suppressed exponentially



# The Wang-Landau method

- Directly calculates density of states r(E) instead of canonical averages
- " Acceptance rate proportional to inverse density of states

$$p_i = \frac{1}{\rho(E_i)}$$
 instead of  $p_i = e^{-E_i/k_BT}$ 

All energy levels visited equally well

$$P(E) = \sum_{c} \delta_{E_{c},E} P(c) \propto \rho(E) \frac{1}{\rho(E)} = 1$$

- Random walk in energy space
  - Flat histogram in energy space => no tunneling problem
- " Free energy accessible

$$F = -k_B T \ln \sum_{E} \rho(E) e^{-E/k_B T}$$

One simulation gives results for all T

# The Wang-Landau algorithm

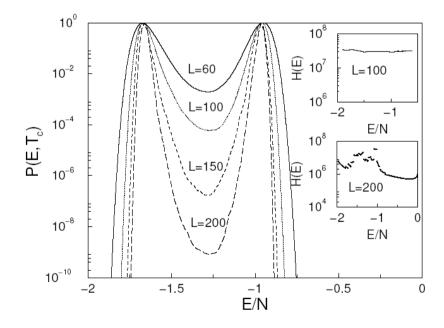
- Initially, r(E) is unknown Start with r(E)=1 and adjust iteratively
- Only a few modifications to usual sampling needed

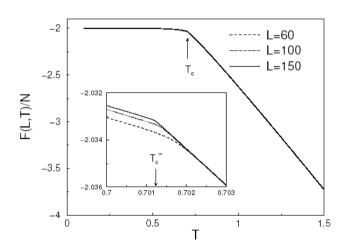
- " Comments
  - Initially: multiplicative changes with large f allow rapid crude convergence
  - Finally: small f means fine structure of r(E) can be mapped out

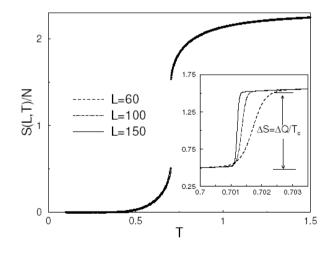
# **Example application of Wang-Landau sampling**

- 10 states Potts model in 2D
  - First order phase transition
  - $T_c = 0.701$

$$P(E,T_c) = \rho(E)e^{-E/k_BT_c}$$







# **Applications and Extensions**

- First and second order phase transitions and disorder
  - F. Wang and D.P. Landau, (2001)
- Scaling analysis for Ising spin glasses
  - P. Dayal, et al., (2003).
- improved sampling and optimized ensembles
  - S. Trebst, D. Huse, and M. Troyer, (2003)
  - C. Zhou and R.N. Bhatt, (2003)
  - B.J. Schulz et al., (2002).
  - C. Yamaguchi and N. Kawashima, (2002).
- Continuum simulations
  - <sup>a</sup> Q. Yan et al., (2002).
  - <sup>\*\*</sup> M.S. Shell et al., (2002).
- " Polymer films
  - T.S. Jan et al., (2002).
- Protein folding
  - <sup>..</sup> N. Rathore et al., (2002).

Try out the Y-Mag Wang-Landau tutorial!

# Quantum version of the Wang-Landau algorithm

Density of states not accessible directly in quantum case

"Classical 
$$Z = \sum_{c} e^{-E_c/k_B T} = \sum_{E} \rho(E) e^{-E/k_B T}$$

- " Quantum  $Z = \operatorname{Tr} e^{-\beta H}$
- Quantum version based on SSE

$$Z = \sum_{n} \sum_{|\alpha\rangle} \sum_{(b_1,\dots,b_\Lambda)} \frac{\beta^n}{n!} \langle \alpha | \prod_{i=1}^{\Lambda} (-H_{b_i}) | \alpha \rangle$$

$$\approx \sum_{n=0}^{\Lambda} \sum_{|\alpha\rangle} \sum_{(b_1,\dots,b_\Lambda)} \frac{(\Lambda - n)! \beta^n}{\Lambda!} \langle \alpha | \prod_{i=1}^{\Lambda} (-H_{b_i}) | \alpha \rangle$$

$$= \sum_{n=0}^{\Lambda} \beta^n g(n)$$

We have a similar expression as a sum over orders n instead of energies

# Quantum version of the Wang-Landau algorithm

- We want flat histogram in order n
  - Use the Wang-Landau algorithm to get

$$Z = \sum_{n=0}^{\Lambda} \beta^{n} g(n) \quad \text{from} \quad Z = \sum_{n=0}^{\Lambda} \sum_{|\alpha\rangle} \sum_{(b_{1},\dots,b_{\Lambda})} \frac{(\Lambda - n)! \beta^{n}}{\Lambda!} \langle \alpha | \prod_{i=1}^{\Lambda} (-H_{b_{i}}) | \alpha \rangle$$

Small change in acceptance rates for diagonal updates

$$P[1 \to H_{(i,j)}^d] = \min \left(1, \frac{\beta N_{bonds} \langle \alpha | H_{(i,j)}^d | \alpha \rangle}{\Lambda - n}\right) \xrightarrow{\text{Wang-Landau}} \min \left(1, \frac{N_{bonds} \langle \alpha | H_{(i,j)}^d | \alpha \rangle}{\Lambda - n}\right) \xrightarrow{\text{g}(n)}$$

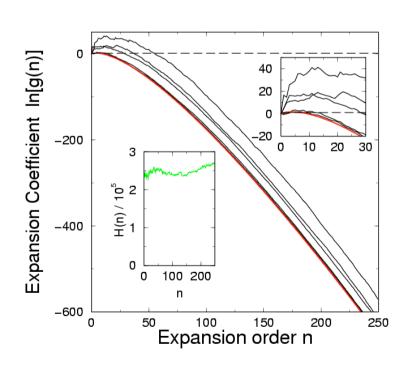
$$P[H_{(i,j)}^{d} \to 1] = \min \left(1, \frac{\Lambda - n + 1}{\beta N_{bonds} \langle \alpha | H_{(i,j)}^{d} | \alpha \rangle}\right) \xrightarrow{\text{Wang-Landau}} \min \left(1, \frac{\Lambda - n + 1}{N_{bonds} \langle \alpha | H_{(i,j)}^{d} | \alpha \rangle} \frac{g(n)}{g(n-1)}\right)$$

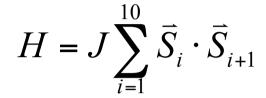
- " Nonlocal updates (directed loops, ...) do not change n and thus remain unchanged!
- Cutoff L limits temperatures to b < L /  $E_0$

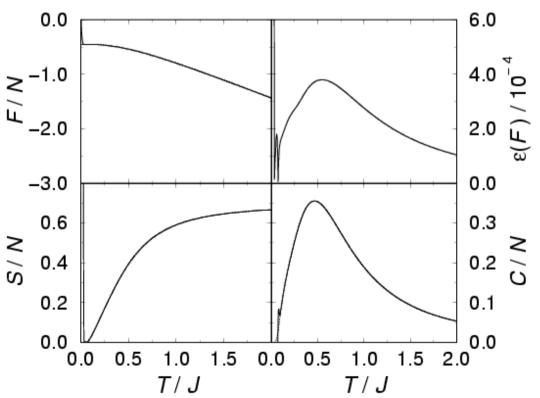
## A first test

- L=10 site Heisenberg chain with L = 250
  - Free energy and entropy available
  - One simulation for all temperatures
  - Clear limit of reliability

T>0.05J due to cutoff

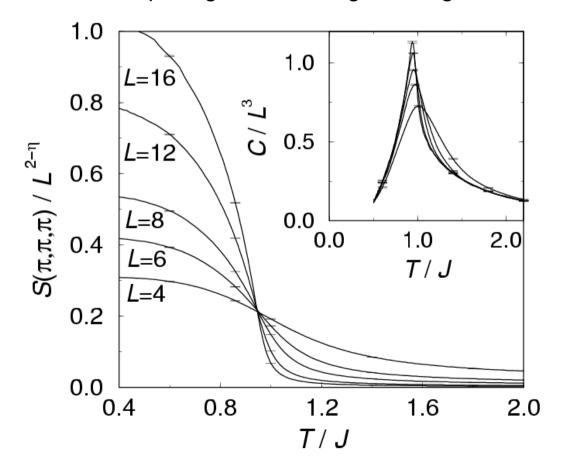






# Thermal phase transitions

- 3D Heisenberg antiferromagnet
  - Second order phase transition at  $T_c$ =0.947J
  - Parallelization via splitting of the *n*-range among several CPUs



# Quantum version – perturbation expansion

Instead of temperature a coupling constant can be varied

$$H = H_0 + \lambda V$$

Based on finite temperature perturbation expansion

$$Z = \operatorname{Tr}(e^{-\beta H}) = \sum_{n=0}^{\infty} \frac{(-\beta)^n}{n!} \sum_{|\alpha\rangle} \sum_{(b_1, \dots, b_n)} \langle \alpha | \prod_{i=1}^n (-H_{b_i}) | \alpha \rangle \lambda^{n_{\lambda}(b_1, \dots, b_n)}$$

$$\approx \sum_{n=0}^{\Lambda} \sum_{|\alpha\rangle} \sum_{(b_1, \dots, b_{\Lambda})} \frac{(\Lambda - n)! \beta^n}{\Lambda!} \langle \alpha | \prod_{i=1}^{\Lambda} (-H_{b_i}) | \alpha \rangle \lambda^{n_{\lambda}(b_1, \dots, b_{\Lambda})}$$

$$= \sum_{n=0}^{\Lambda} \lambda^{n_{\lambda}} g(n_{\lambda})$$

$$n_{\lambda}(b_1, \dots, b_n) \text{ counts } \# \text{ of } \lambda \text{ terms}$$

Flat histogram in order  $n_I$  of perturbation expansion

# Perturbation series by Wang-Landau

- We want flat histogram in order  $n_I$ 
  - Use the Wang-Landau algorithm to get

$$Z = \sum_{n_{\lambda}=0}^{\Lambda} \lambda^{n_{\lambda}} g(n_{\lambda}) \quad \text{from} \quad Z = \sum_{n=0}^{\Lambda} \sum_{|\alpha\rangle} \sum_{(b_{1},\dots,b_{\Lambda})} \frac{(\Lambda - n)! \beta^{n}}{\Lambda!} \langle \alpha | \prod_{i=1}^{\Lambda} (-H_{b_{i}}) | \alpha \rangle \lambda^{n_{\lambda}(b_{1},\dots,b_{\Lambda})}$$

Small change in acceptance rates for diagonal updates

$$P[1 \to H_{(i,j)}^d] = \min \left(1, \frac{\beta N_{bonds} \langle \alpha | H_{(i,j)}^d | \alpha \rangle}{\Lambda - n}\right) \xrightarrow{\text{Wang-Landau}} \min \left(1, \frac{\beta N_{bonds} \langle \alpha | H_{(i,j)}^d | \alpha \rangle}{\Lambda - n}\right) \xrightarrow{g(n_{\lambda})} \frac{g(n_{\lambda})}{g(n_{\lambda} + \Delta n_{\lambda})}$$

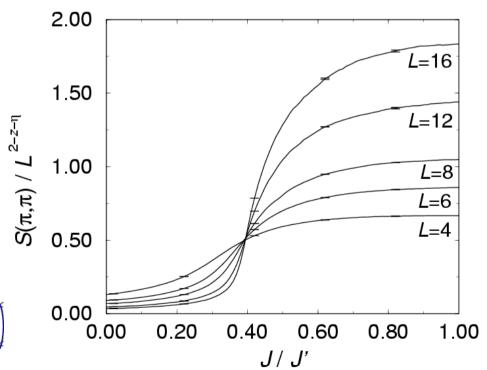
$$P[H_{(i,j)}^{d} \to 1] = \min \left(1, \frac{\Lambda - n + 1}{\beta N_{bonds} \langle \alpha | H_{(i,j)}^{d} | \alpha \rangle}\right) \xrightarrow{\text{Wang-Landau}} \min \left(1, \frac{\Lambda - n + 1}{\beta N_{bonds} \langle \alpha | H_{(i,j)}^{d} | \alpha \rangle} \frac{g(n_{\lambda})}{g(n_{\lambda} - \Delta n_{\lambda})}\right)$$

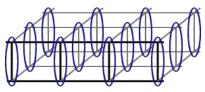
- " Nonlocal updates (directed loops, ...) do not change  $n_{\lambda}$  and thus remain unchanged!
- Cutoff L limits value of I for which the series converges

# Quantum phase transition

- Bilayer Heisenberg Antiferromagnet
  - Single simulation gives results for all values of  $\lambda = J/J'$
  - " Quantum phase transition at  $\lambda = 0.396$
  - " Parallelization as in thermal case

$$H = J \sum_{p=1}^{2} \sum_{\langle i,j \rangle} \vec{S}_{i,p} \cdot \vec{S}_{j,p} + J' \sum_{i} \vec{S}_{i,1} \cdot \vec{S}_{i,2}$$

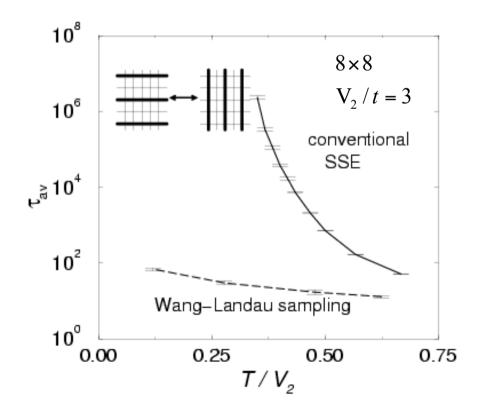


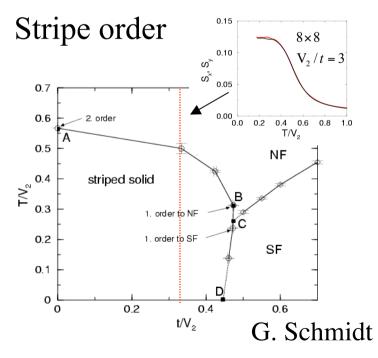


# Speedup at first order phase transitions

- Greatly reduced tunneling times at free energy barriers
  - 2D hard-core bosons

$$H = -t \sum_{\langle i,j \rangle} \left( a_i^{\dagger} a_j + h.c. \right) + V_2 \sum_{\langle \langle i,j \rangle \rangle} n_i n_j$$





# **Summary**

- Extension of Wang-Landau sampling to quantum systems
- Stochastically evaluate series expansion coefficients
  - High-temperature series

$$Z = \sum_{n=0}^{\infty} \frac{\beta^n}{n!} g(n)$$

Perturbation series

$$Z = \sum_{n_{\lambda}=0}^{\infty} \lambda^{n_{\lambda}} g(n_{\lambda})$$

- Features
  - Flat histogram in the expansion order
  - Allows calculation of free energy
  - Like classical systems, allows tunneling through free energy barriers

# Qantum Wang-Landau in ALPS

- Application qmc/qwl
- Allows high temperature expansions for isotropic spin-1/2 systems
- You can reproduce the data presented here in the *qwl tutorial*.

#### Tutorial V: Quantum Wang-Landau Monte Carlo Simulations

- Tutorial I: Running and evaluating MC simulations using ALPS
- Tutorial II: Classical Monte Carlo Simulations
- Tutorial III: Quantum Monte Carlo Simulations (SSE)
- Tutorial IV: Quantum Monte Carlo Simulations (worm code)
- Tutorial V: Quantum Wang-Landau Monte Carlo Simulations
- . Tutorial VI: Density Matrix Renormalization Group for a particle in a box (non-interacting DMRG)

#### Part I: Thermodynamics of Heisenberg quantum spin chains

#### The ferromagnetic Heisenberg chain

The parameter file **qwl/parm1** sets up a Monte Carlo simulation of the quantum mechanical Heisenberg **ferromagnet** on a **one-dimensional chain** with 40 sites, using the quantum Wang-Landau (QWL) method.

```
LATTICE="chain lattice"

MODEL="spin"

LATTICE_LIBRARY="../lattices.xml"

MODEL_LIBRARY="../models.xml"

S = 1/2
L = 40

T_MIN = 0.1

T_MAX = 10.0

DELTA_T = 0.1

CUTOFF = 500

{J = -1}
```