PARALLEL TEMPERING CLUSTER ALGORITHM FOR COMPUTER SIMULATIONS OF CRITICAL PHENOMENA

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 provides an efficient method to investigate systems with rugged free-energy landscapes, particularly at low temperatures

*used in many disciplines:

- biomolecules
- bioinformatics
- classical and quantum frustrated spin system
- •QCD
- •spin glasses
- zeolite structure solution

How it works?

different replica are simulated at different temperatures
regular intervals an attempt is made to exchange the replica
replica are exchanged via a Monte Carlo process the attempt is accepted with a probability

 $P_{\rm PT}(E_1, \beta_1 \to E_2, \beta_2) = \min[1, \exp(\Delta\beta\Delta E)]$

with $\Delta\beta = \beta_2 - \beta_1$ and $\Delta E = E_2 - E_1$

How it works?



 T_1

T₂

 T_3

 T_4

How it works?



 T_1

 T_2

 T_3

 T_4







$T_{1} \qquad T_{1} \qquad T_{1} \qquad T_{2} \qquad T_{2$

$P_{\rm PT}(E_1, \beta_1 \to E_2, \beta_2) = \min[1, \exp(\Delta\beta\Delta E)]$

An efficient selection of the temperature intervals for PT simulations is still an open problem.

Several strategies have been proposed:

- based on the assumption of constant overlap between the replica
- based on the maximum flow in the temperature space

Following the concept of constant acceptance rate between replica:

$$A(1 \to 2) = \sum_{E_1, E_2} P_{\beta_1}(E_1) P_{\beta_2}(E_2) P_{\text{PT}}(E_1, \beta_1 \to E_2, \beta_2),$$

where $P_{\beta_i}(E_i)$ is the probability for replica *i* with β_i to have the energy E_i .



Energy distributions of the 2D Ising model with L = 16 for a set of inverse temperatures starting $\beta_i = 0.38$ and $A(i \rightarrow i + 1) = 0.25$ and 0.5.

[P. Beale, Phys. Rev. Lett. 76, 78 (1996)]

Autocorrelation times



Autocorrelation times τ as a function β of for the independent simulations and the parallel tempering update scheme for the 2D Ising model (L = 80).

*cover the complete desired "critical" temperature range



$$C(\beta) = \beta^2 V(\langle e^2 \rangle - \langle e \rangle^2)$$

$$\chi(\beta) = \beta V(\langle m^2 \rangle - \langle |m| \rangle^2)$$

$$U_{2k}(\beta) = 1 - \langle m^{2k} \rangle / 3 \langle |m|^k \rangle^2$$

*cover the complete desired "critical" temperature range



$$S = \{C, \chi, \dots\}$$

$$S^{\max} = S(\beta_S^{\max})$$

$$S(\beta_S^{+/-}) = rS^{\max}$$

$$\beta_S^+ > \beta_S^{\max} \text{ and } \beta_S^- < \beta_S^{\max}$$
''desired'' range: $[\beta_{S_{k_1}}^-, \beta_C^+]$

*cover the complete desired "critical" temperature range



$$S = \{C, \chi, \dots\} \qquad r = \frac{2}{3}$$

$$S^{\max} = S(\beta_S^{\max}) \qquad r = \frac{2}{3}$$

$$S(\beta_S^{+/-}) = nS^{\max}$$

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general recipe:

- I. compute the simulation temperatures of the replica equidistant in β ,
- 2. perform several hundred thermalization sweeps and a short measurement run,
- 3. check the histogram overlap between adjacent replica: if the overlap is too small, add on or two replica and goto step 1, else go on,
- 4. use multi-histogram reweighting to determine β_S^- and β_S^+ for all observables *S*,
- 5. leading to the temperature interval $[\beta_{\min}^-, \beta_{\max}^+] = [\min_{S} \{\beta_S^-\}, \max_{S} \{\beta_S^+\}],$
- 6. start with $\beta^- = \beta_{\min}^-$ and compute a sequence of temperatures β_i with fixed acceptance rate $A(1 \rightarrow 2)$ until $\beta_i = \beta^+ \ge \beta_{\max}^+$,
- 7. perform several hundred thermalization sweeps and a long measurement run.



The 'desired' temperature interval for r = 2/3 as a function of the system size.

Autocorrelation times



Autocorrelation times τ_{int} and τ_{eff} for the energy of the 2D Ising model, where $\tau_{eff} = N_{rep}\tau_{int}$ and N_{rep} is the number of replica.

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understand the FSS of τ_{eff} $\Delta\beta_{\text{rew}} \propto L^{-1/\nu}/\sqrt{\ln L}$ $\beta^+ - \beta^- \approx \beta_C^+ - \beta_{S_{k_1}}^- = aL^{-1/\nu} + bL^{-r/\nu}$ $N_{\text{rep}} = (\beta^+ - \beta^-)/\Delta\beta_{\text{rew}} \rightarrow L^{(1-r)/\nu}\sqrt{\ln L}$ $\beta^+ - \beta^- \propto L^{-\kappa'}$

in 2D Ising



64x64

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in 2D Ising





FSS of the "desired" simulation window for the 2D Ising model with r = 2/3.

Autocorrelation times



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What can we do to improve the parallel tempering algorithm?

•use the replica-exchange cluster algorithm or

- •use a constant acceptance rate between the replica
- keep the temperatures fixed
- take the temperature dependence of autocorrelation times into account

EB, A. Nußbaumer, and W. Janke, Phys. Rev. Lett. 101 (2008) 130603 EB and W. Janke, Phys. Rev. E 84 (2011) 036701

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THANKYOU!



The way through inverse temperature space of an arbitrarily chosen replica:



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$\begin{array}{l} \text{2D Ising model} \\ (L = 80) \end{array}$

The flow η is the fraction of replica which wander from the largest β to the smallest as a function of the replica index *i*.

$$\eta = \frac{n_{\rm up}(i)}{n_{\rm up}(i) + n_{\rm down}(i)}$$



[H.G. Katzgraber, S. Trebst, D.A. Huse, and M. Troyer, J. Stat. Mech. P03018 (2006)]

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flow for the 2D Ising model (L = 80)

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flow for the 2D Ising model (L = 80)

Autocorrelation times



Autocorrelation times τ as a function β of for the independent simulations and the parallel tempering update scheme for the 2D Ising model (L = 80). Improved parallel tempering update scheme



 $N_{\rm local}(\beta) \propto \tau_{\rm can}(\beta)$

Autocorrelation times



Autocorrelation times τ as a function β of for the independent simulations and the parallel tempering update scheme for the 2D Ising model (L = 80).

Autocorrelation times



Autocorrelation times auas a function β of for the independent simulations, the parallel tempering update scheme, the feedback-optimized parallel tempering method, and the improved parallel tempering update scheme for the 2D Ising model(L = 80).

The flow η is the fraction of replica which wander from the largest β to the smallest as a function of the replica index *i*.

$$\eta = \frac{n_{\rm up}(i)}{n_{\rm up}(i) + n_{\rm down}(i)}$$



flow for the 2D Ising model (L = 80)

Sweeps per tunneling



Sweeps per tunneling as a function of N_{local} for the 2D Ising model (L = 80).