

# PARALLEL TEMPERING CLUSTER ALGORITHM FOR COMPUTER SIMULATIONS OF CRITICAL PHENOMENA

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# Parallel Tempering

- ❖ 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

# Parallel Tempering

## ❖ 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

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

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

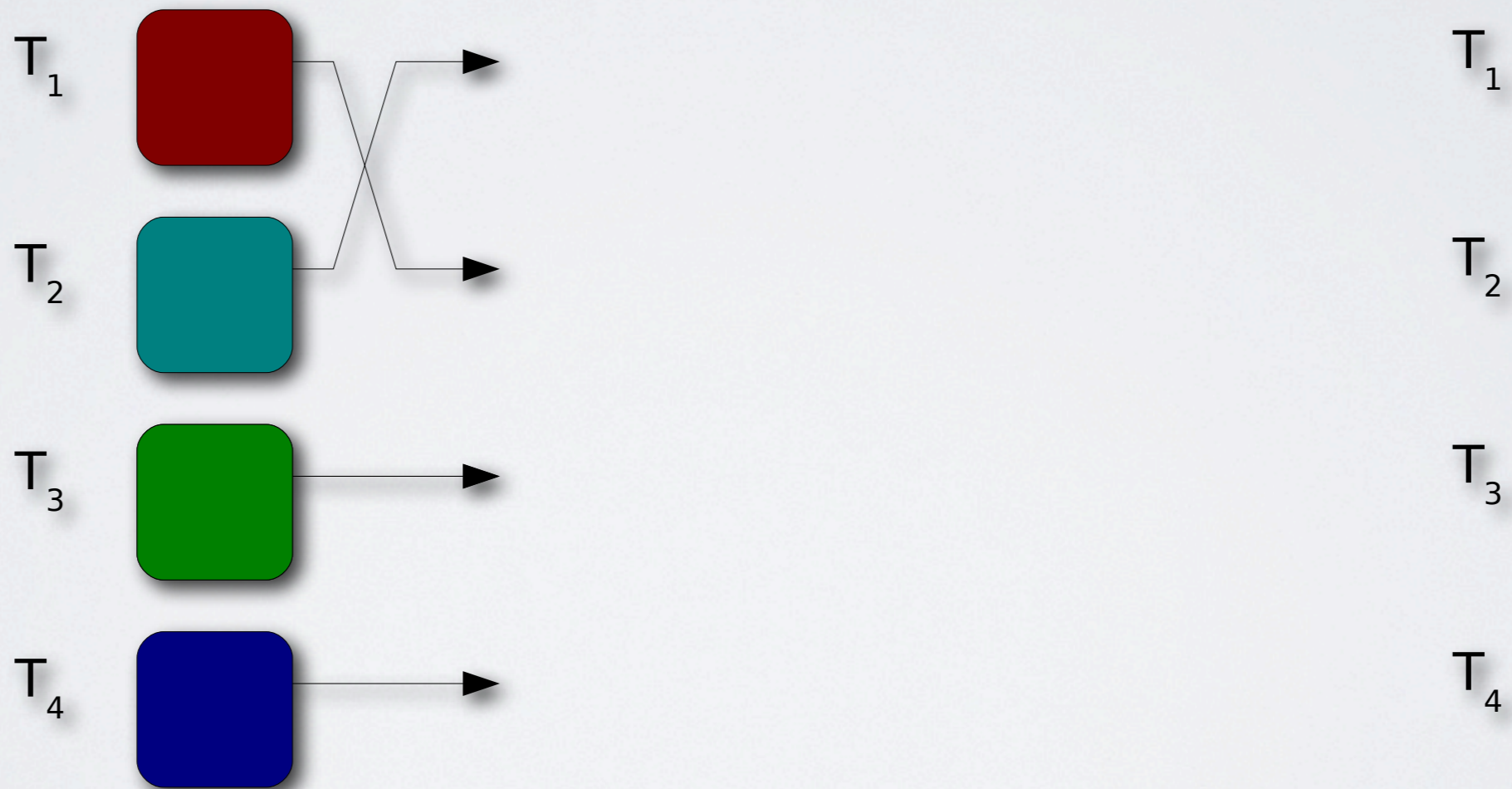
# Parallel Tempering

❖ How it works?



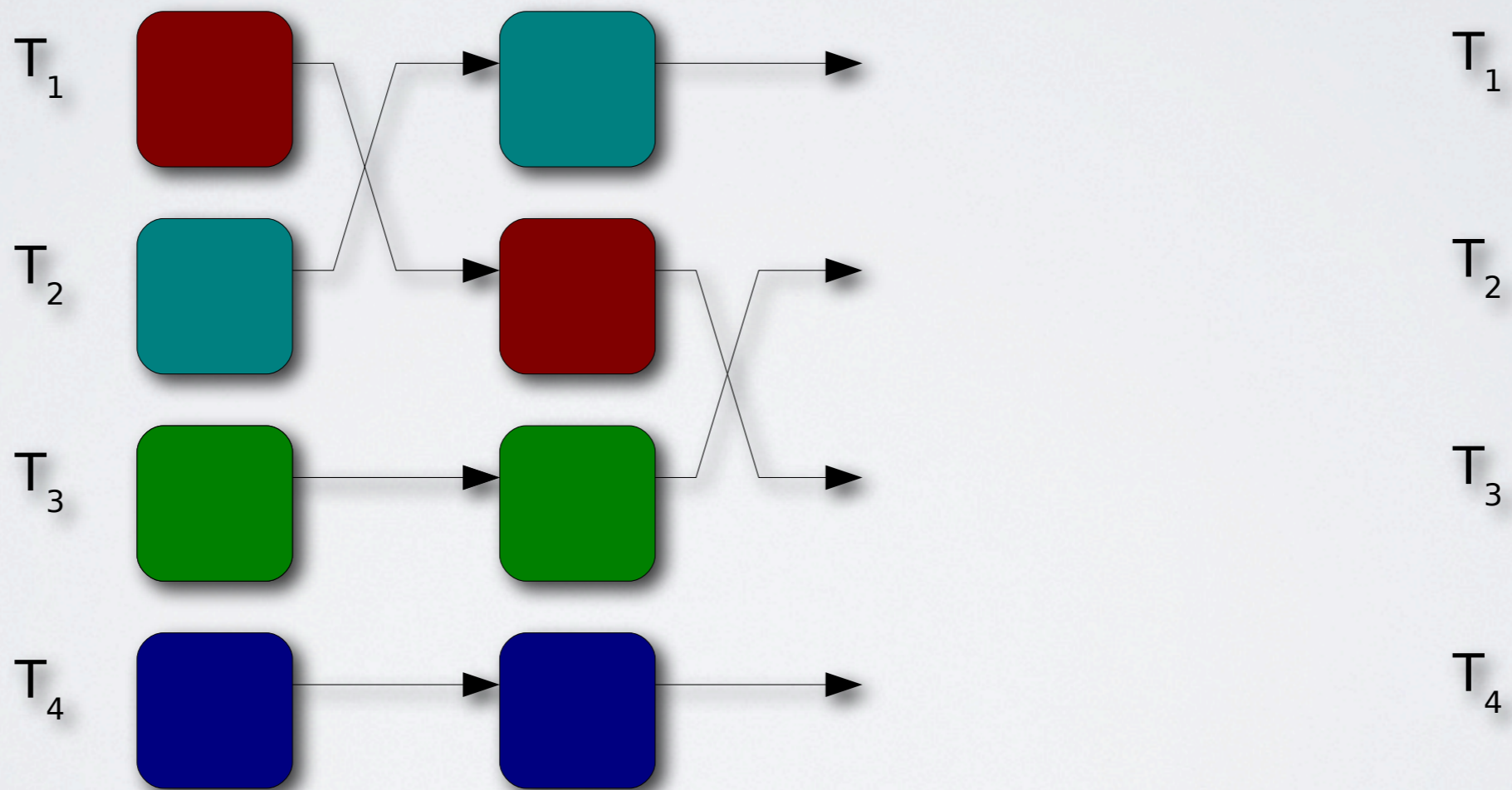
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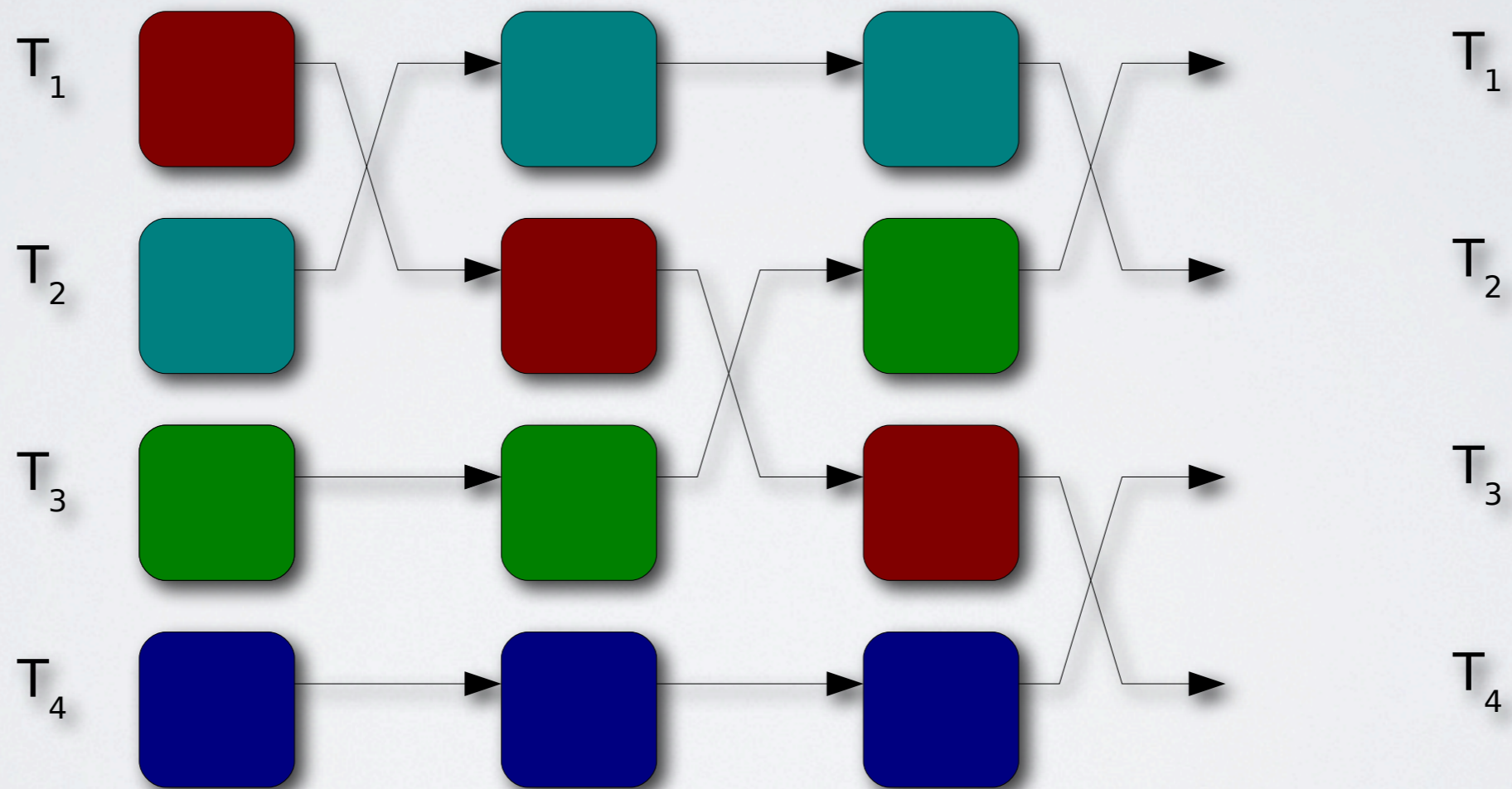
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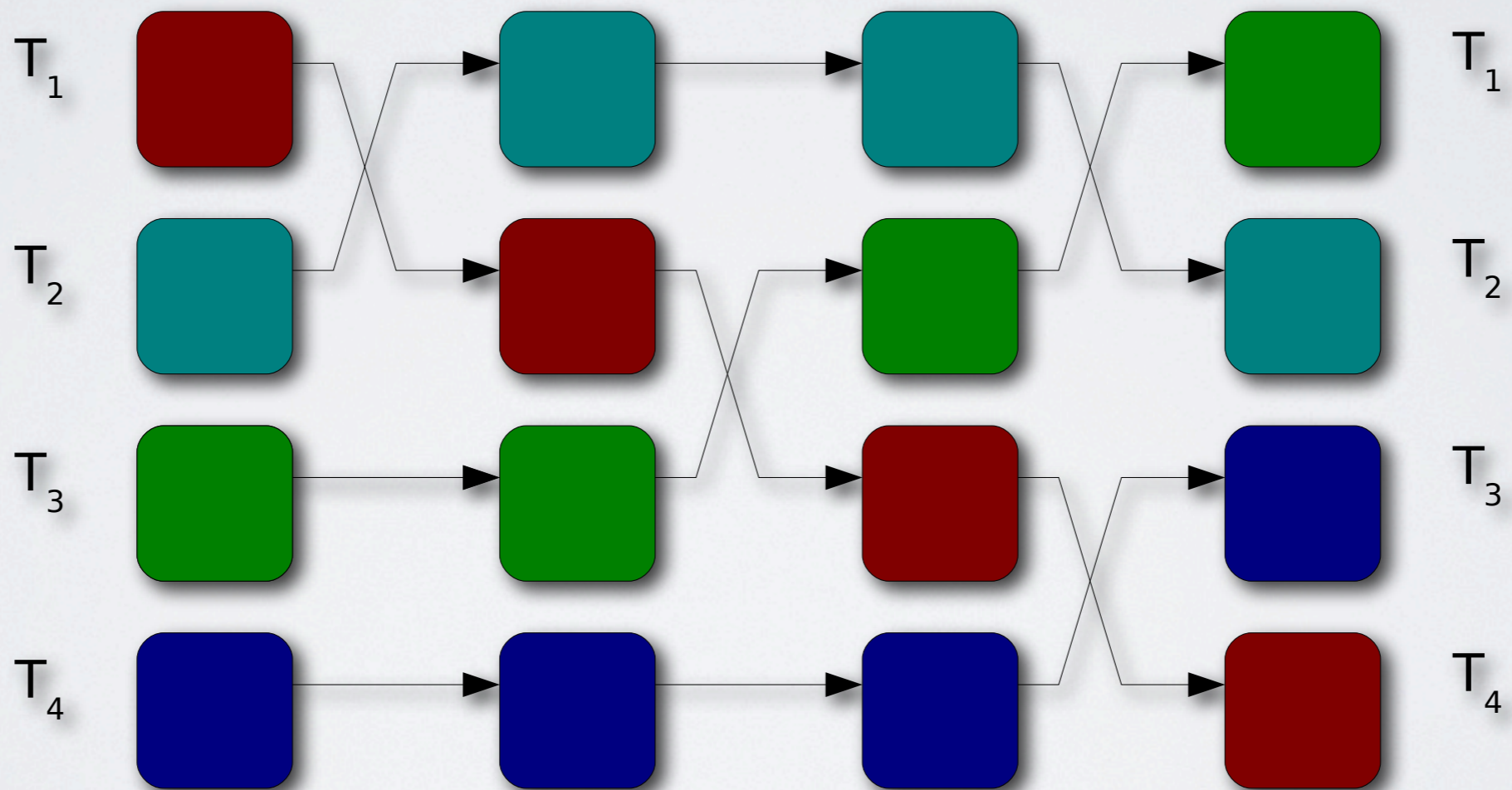
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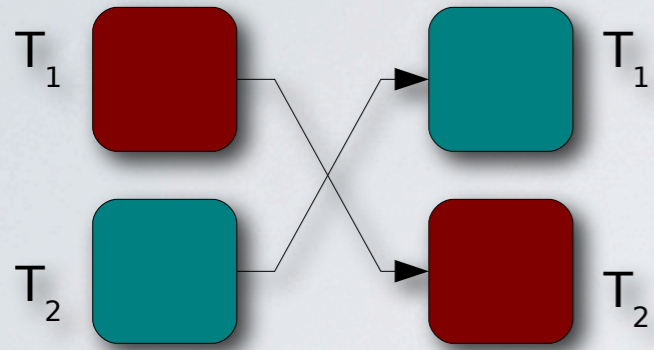
# Parallel Tempering

❖ How it works?





# Parallel Tempering



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

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

Several strategies have been proposed:

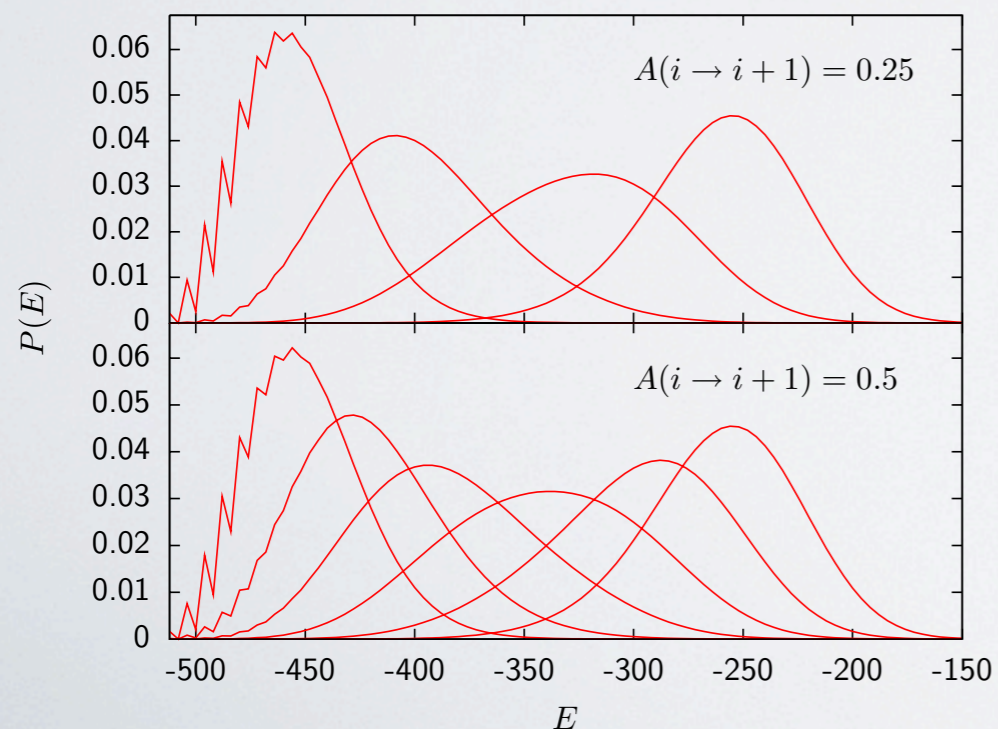
- based on the assumption of constant overlap between the replica
- based on the maximum flow in the temperature space

# Parallel Tempering

Following the concept of constant acceptance rate between replica:

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

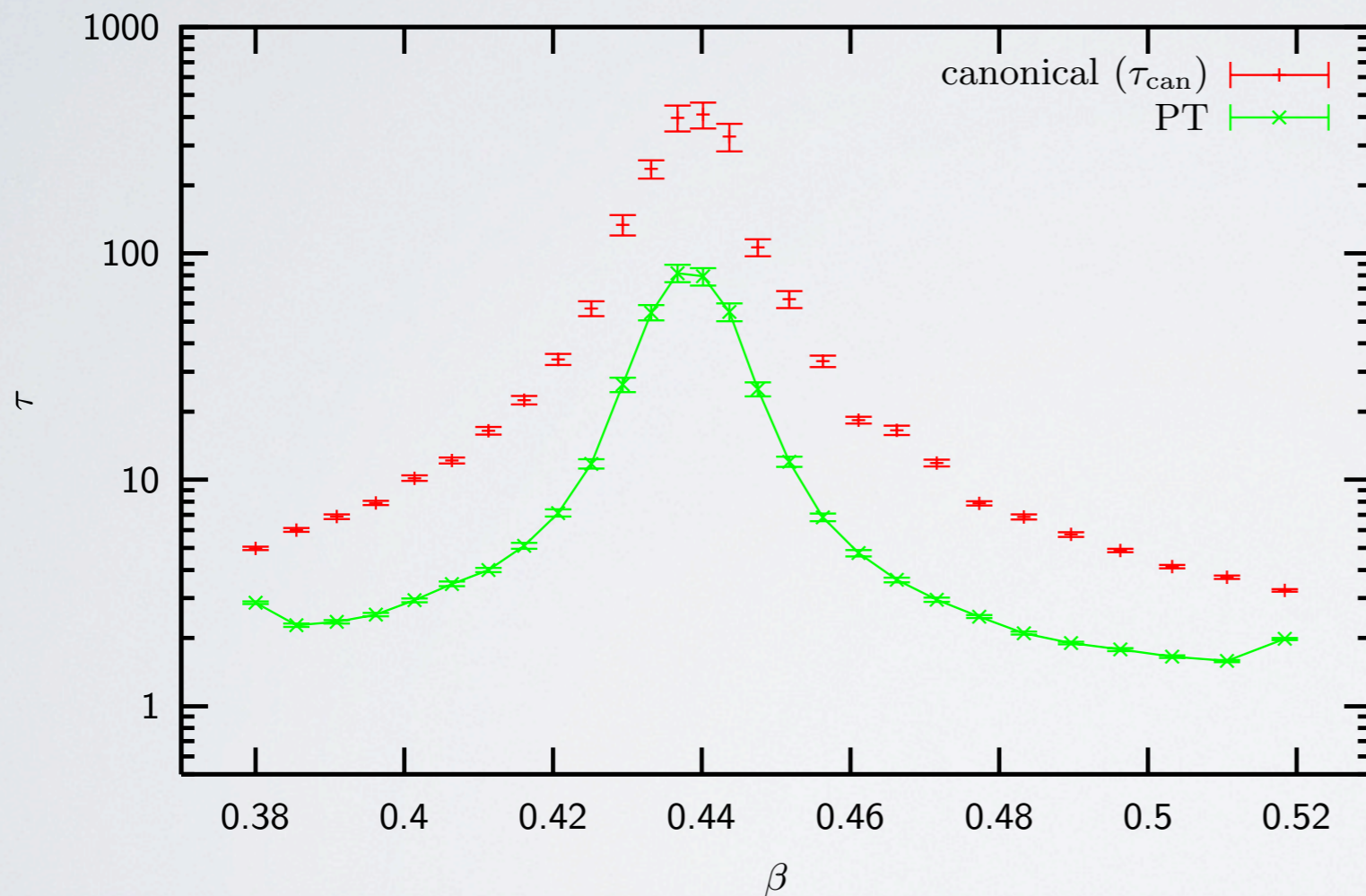
where  $P_{\beta_i}(E_i)$  is the probability for replica  $i$  with  $\beta_i$  to have the energy  $E_i$ .



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)]

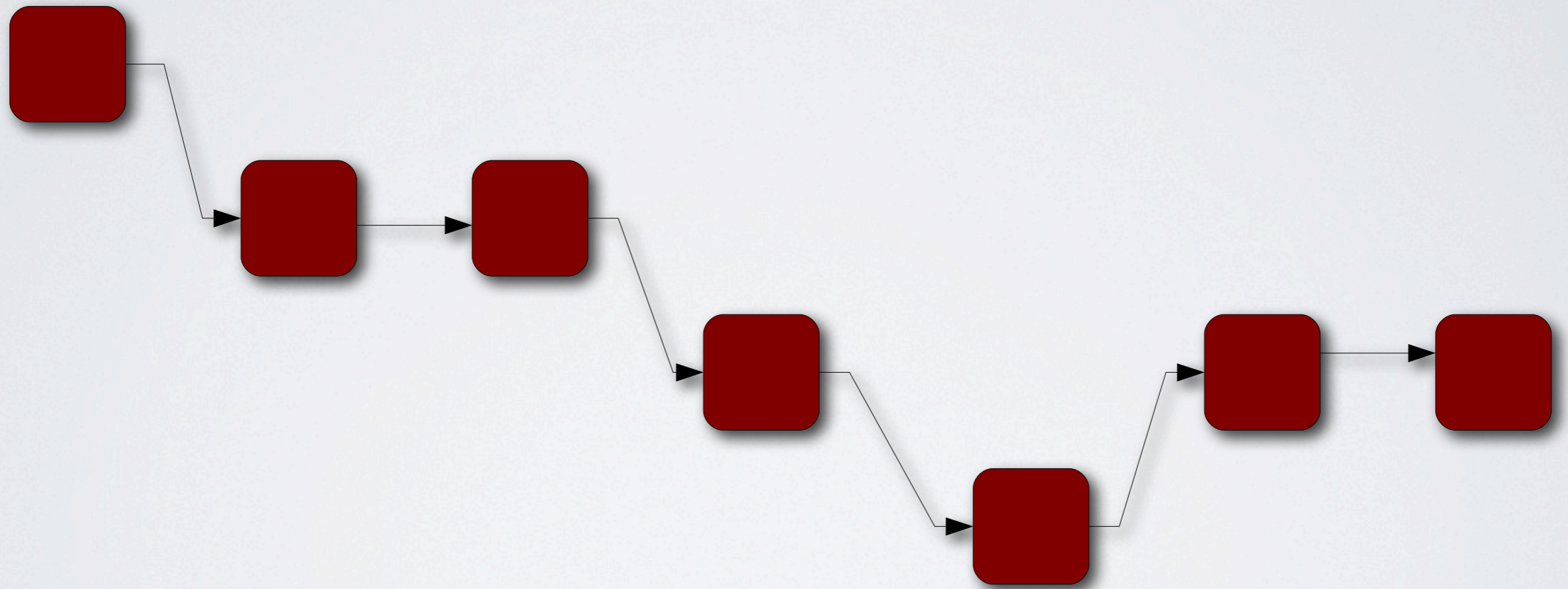
# Autocorrelation times



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

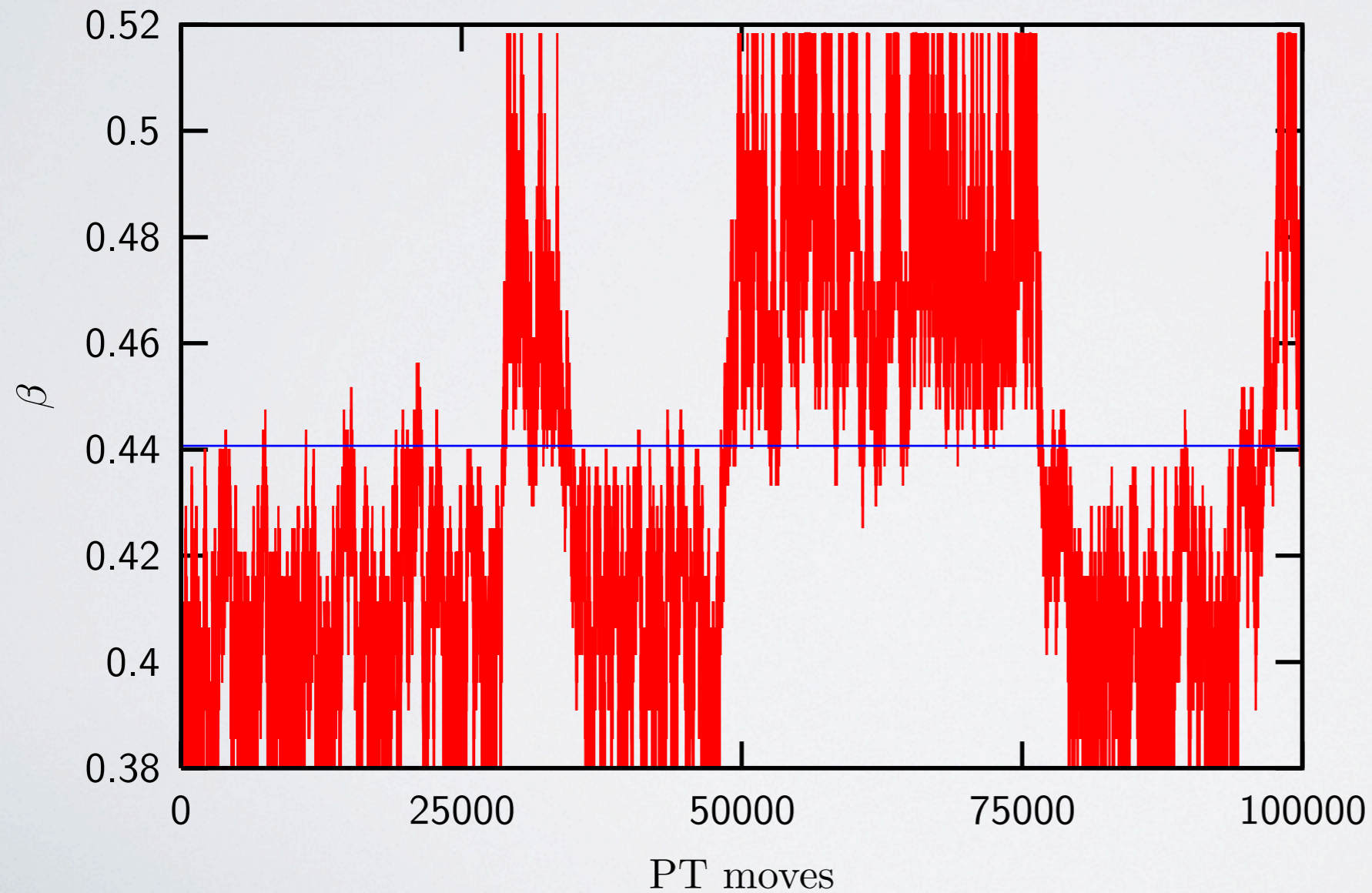
# Flow

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



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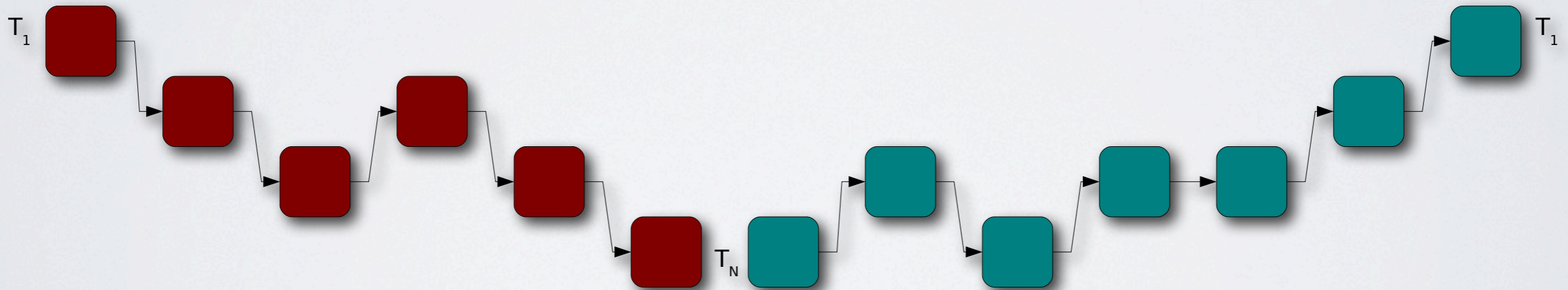


2D Ising model  
( $L = 80$ )

# Flow

The flow  $\eta$  is the fraction of replica which wander from the largest  $\beta$  to the smallest as a function of the replica index  $i$ .

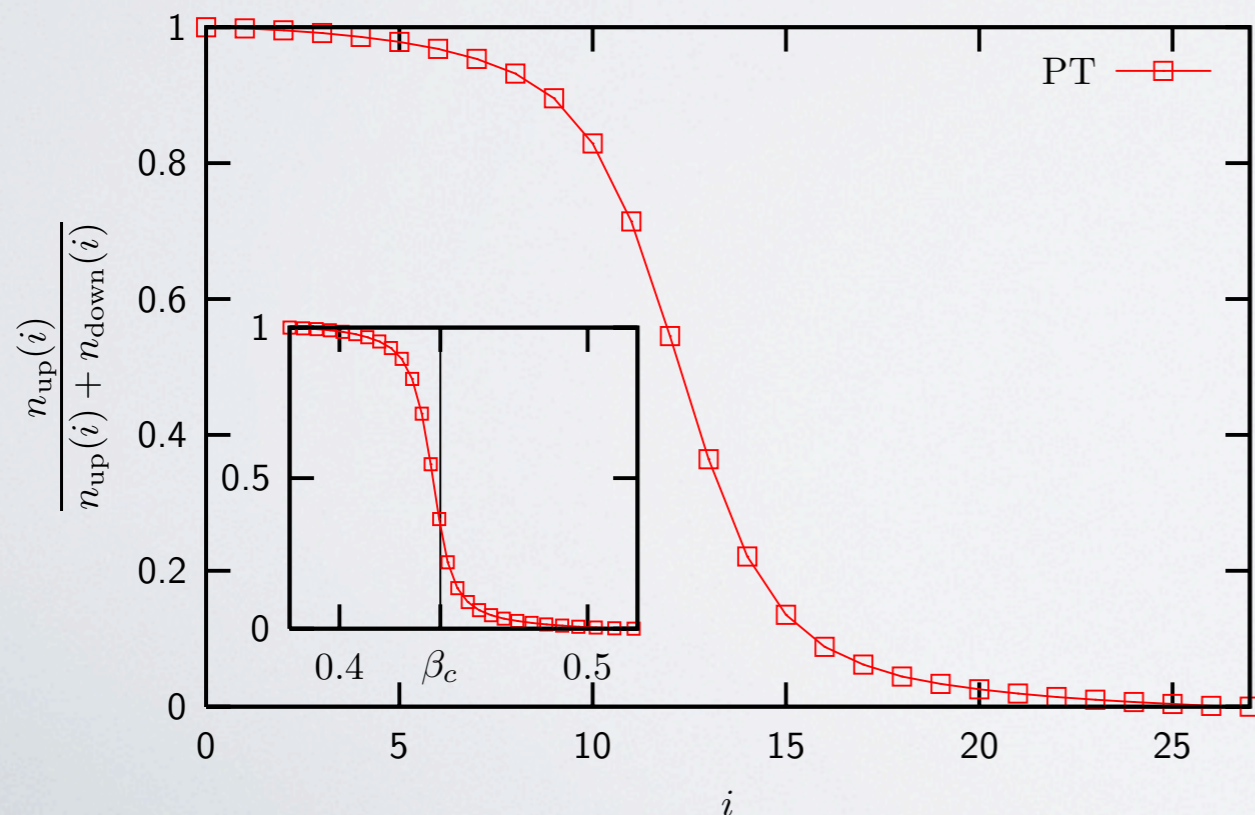
$$\eta = \frac{n_{\text{up}}(i)}{n_{\text{up}}(i) + n_{\text{down}}(i)}$$



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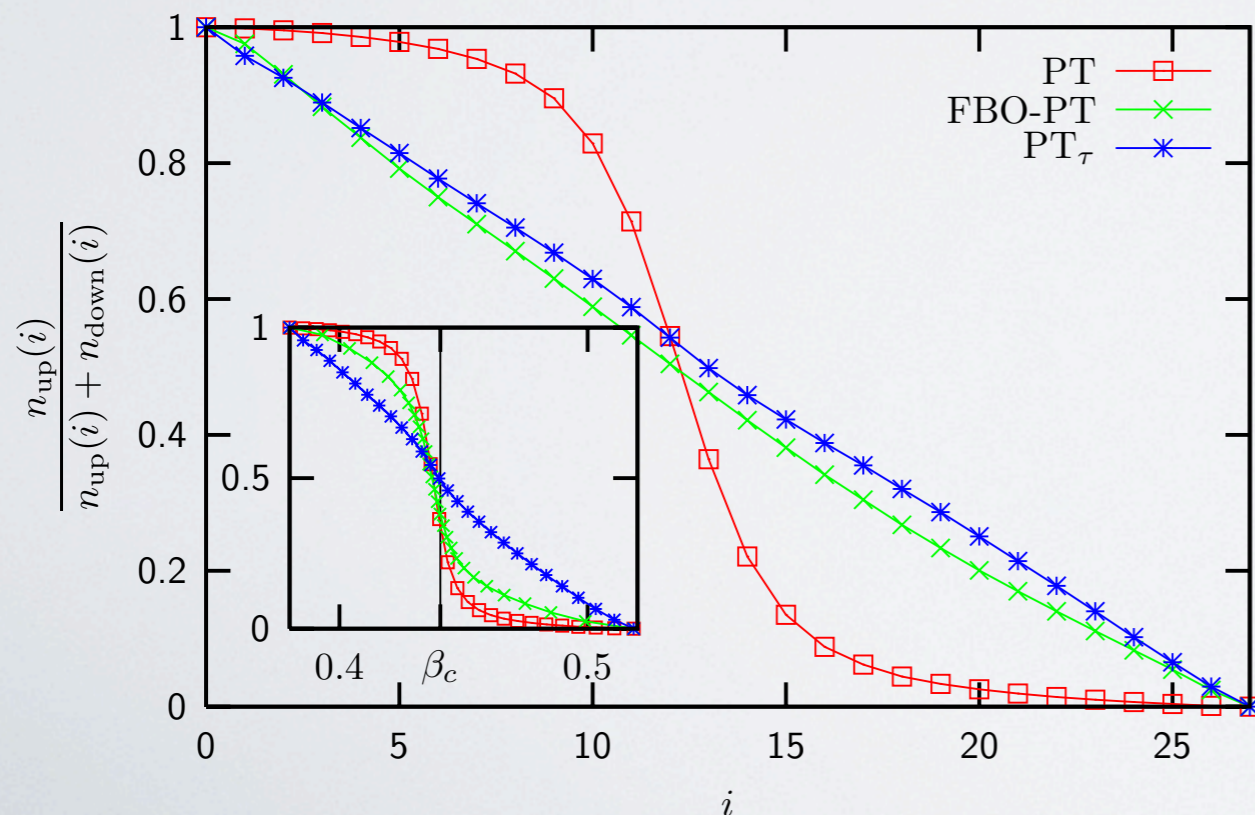


flow for the 2D Ising model ( $L = 80$ )

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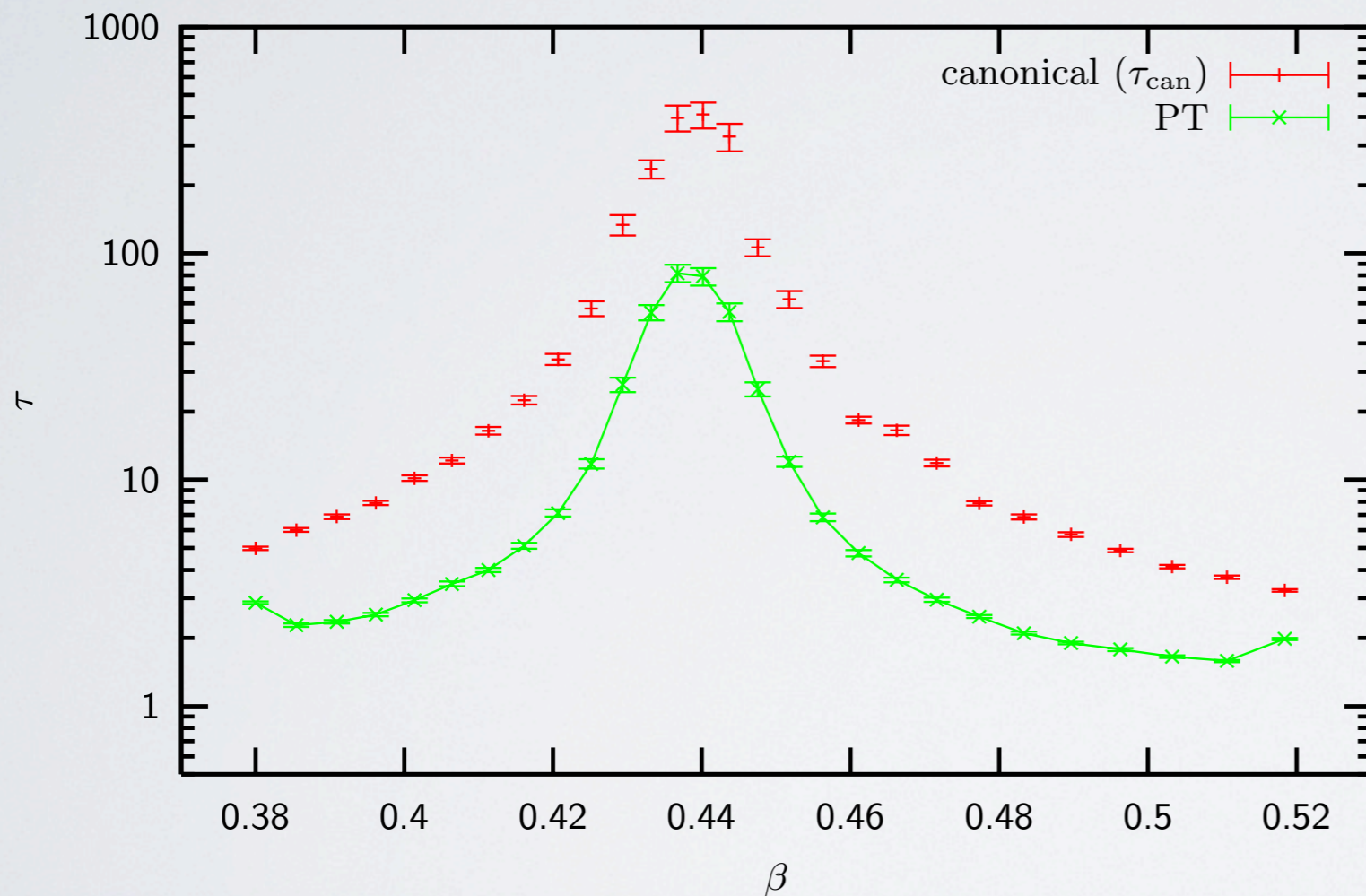
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flow for the 2D Ising  
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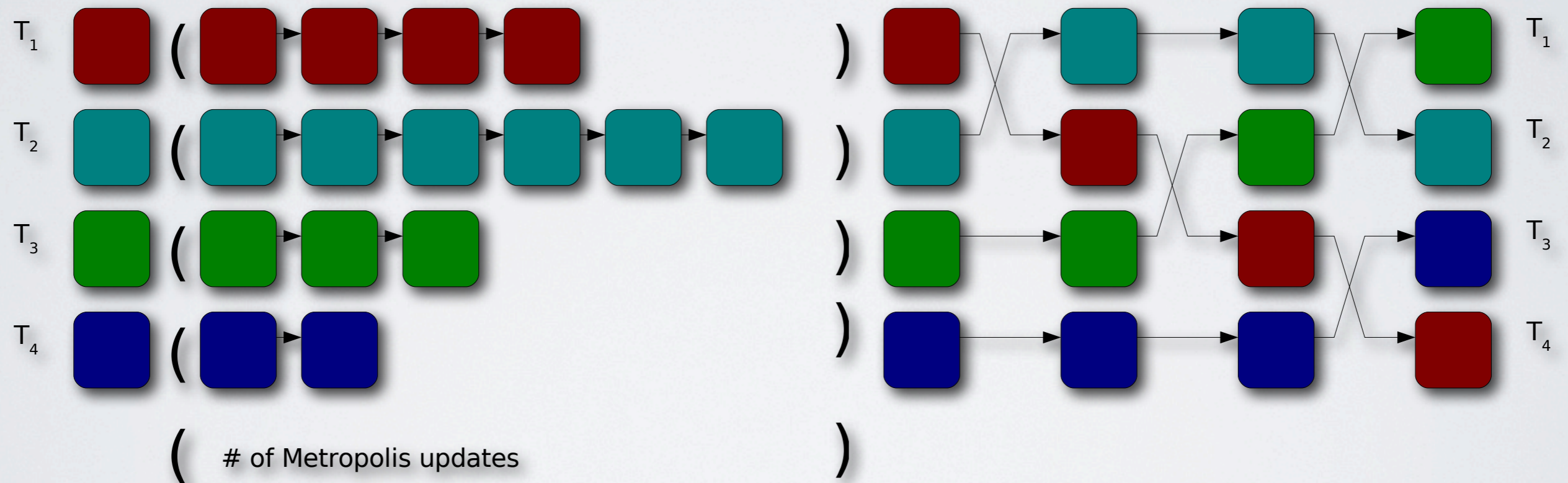
# Autocorrelation times



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

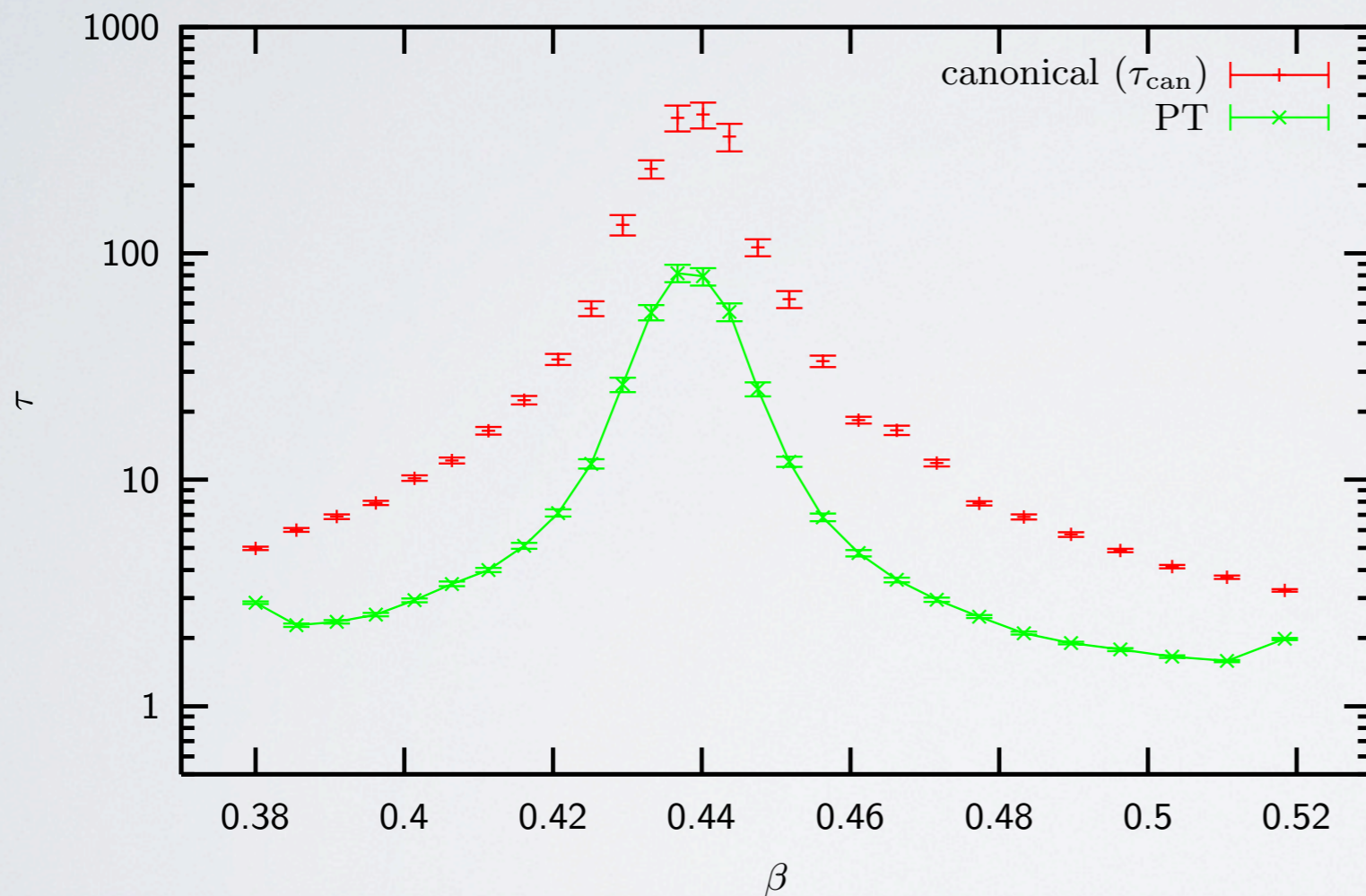
# Improved parallel tempering update scheme

❖ How it works?



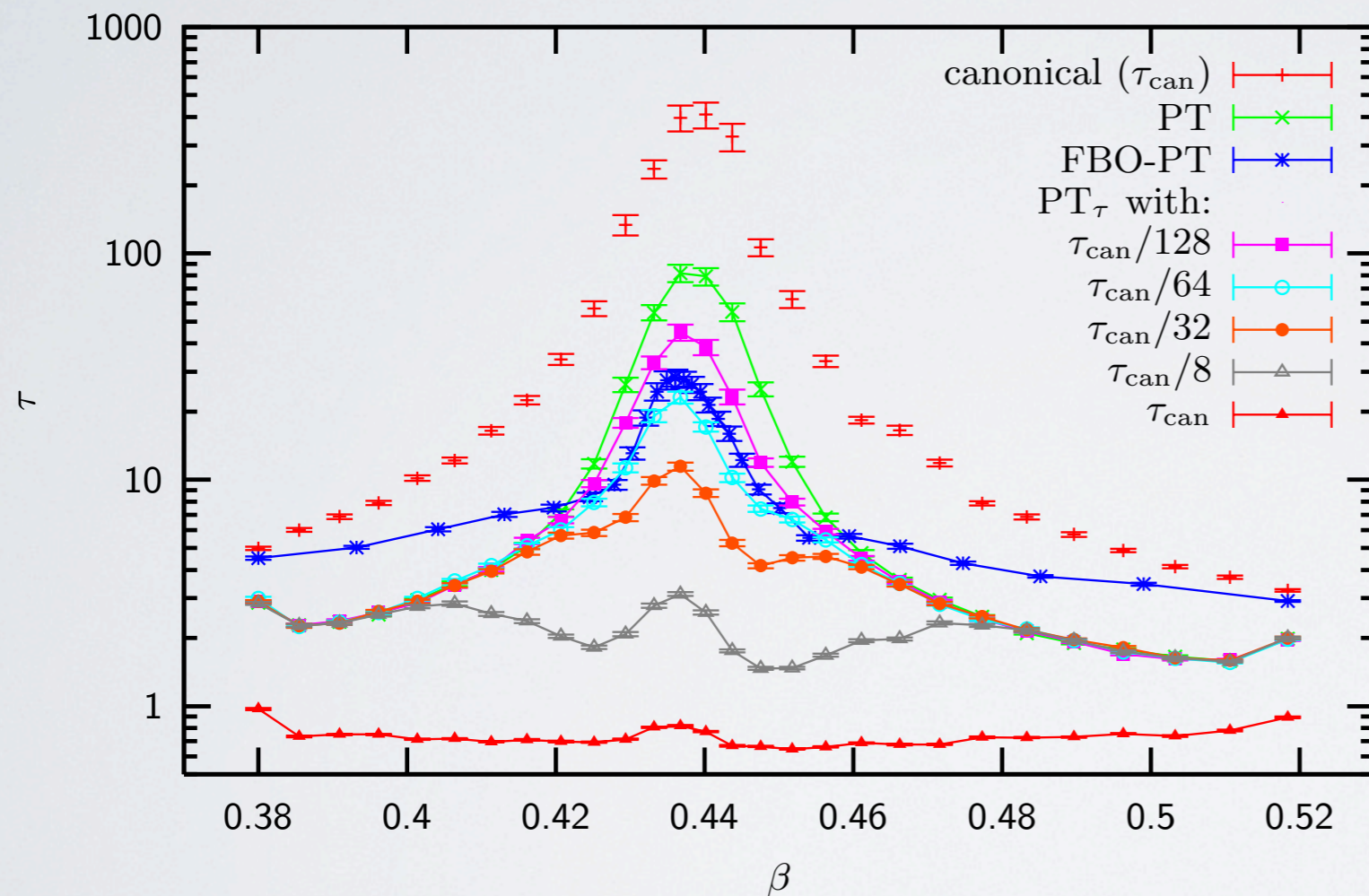
$$N_{\text{local}}(\beta) \propto \tau_{\text{can}}(\beta)$$

# Autocorrelation times



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

# Autocorrelation times

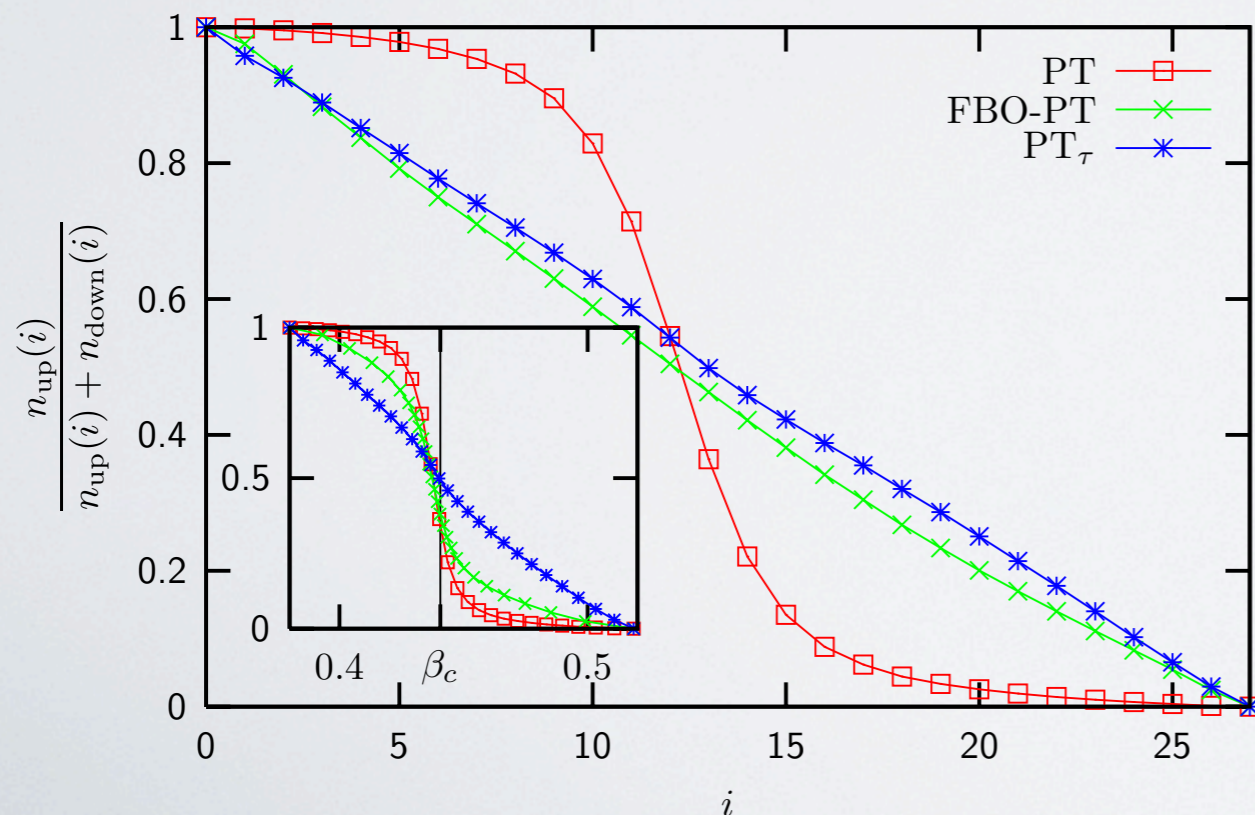


Autocorrelation times  $\tau$  as a function  $\beta$  of for the independent simulations, the parallel tempering update scheme, the feedback-optimized parallel tempering method, and the improved parallel tempering update scheme for the 2D Ising model ( $L = 80$ ).

# Flow

The flow  $\eta$  is the fraction of replica which wander from the largest  $\beta$  to the smallest as a function of the replica index  $i$ .

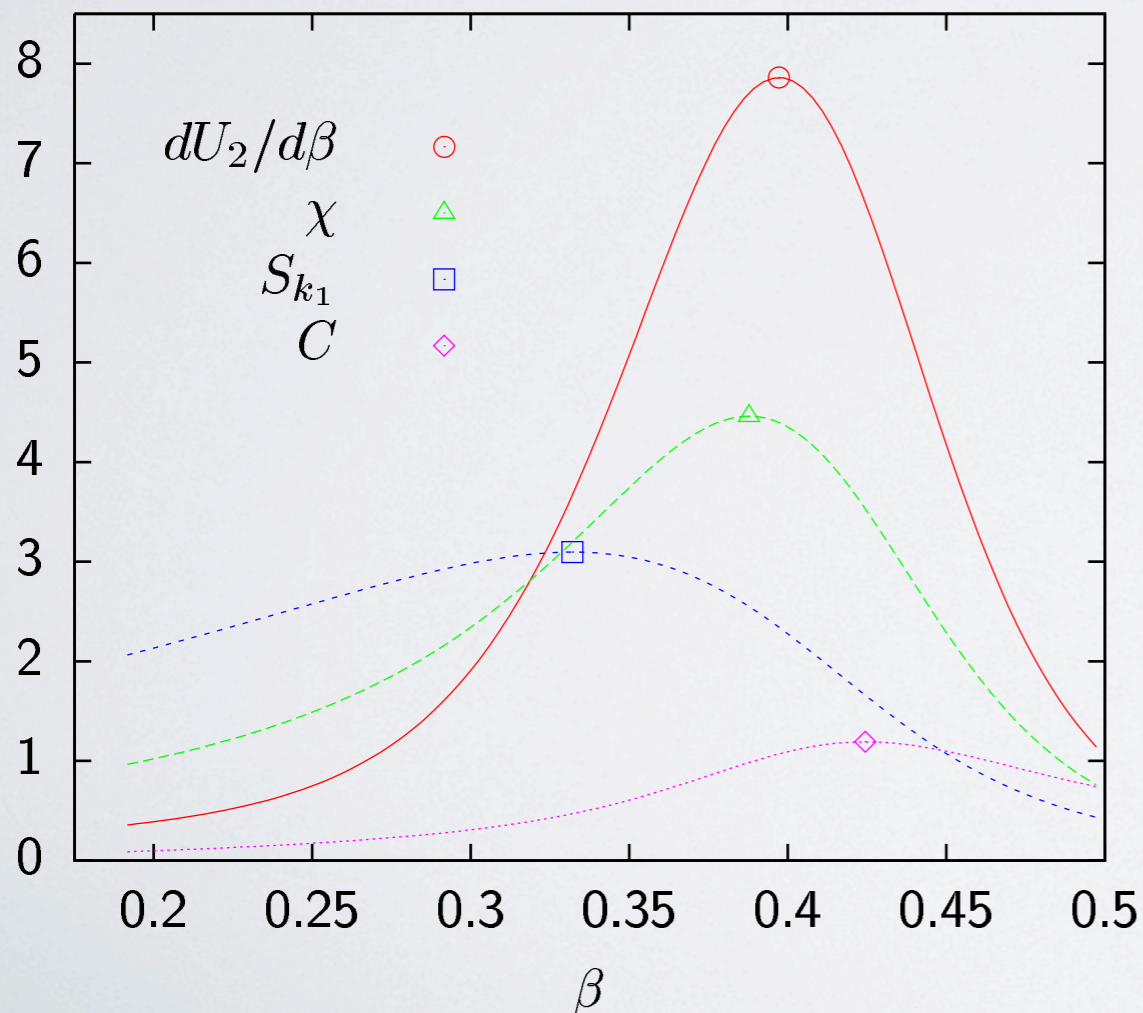
$$\eta = \frac{n_{\text{up}}(i)}{n_{\text{up}}(i) + n_{\text{down}}(i)}$$



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

# Temperature Interval

❖ cover the complete desired “critical” temperature range



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

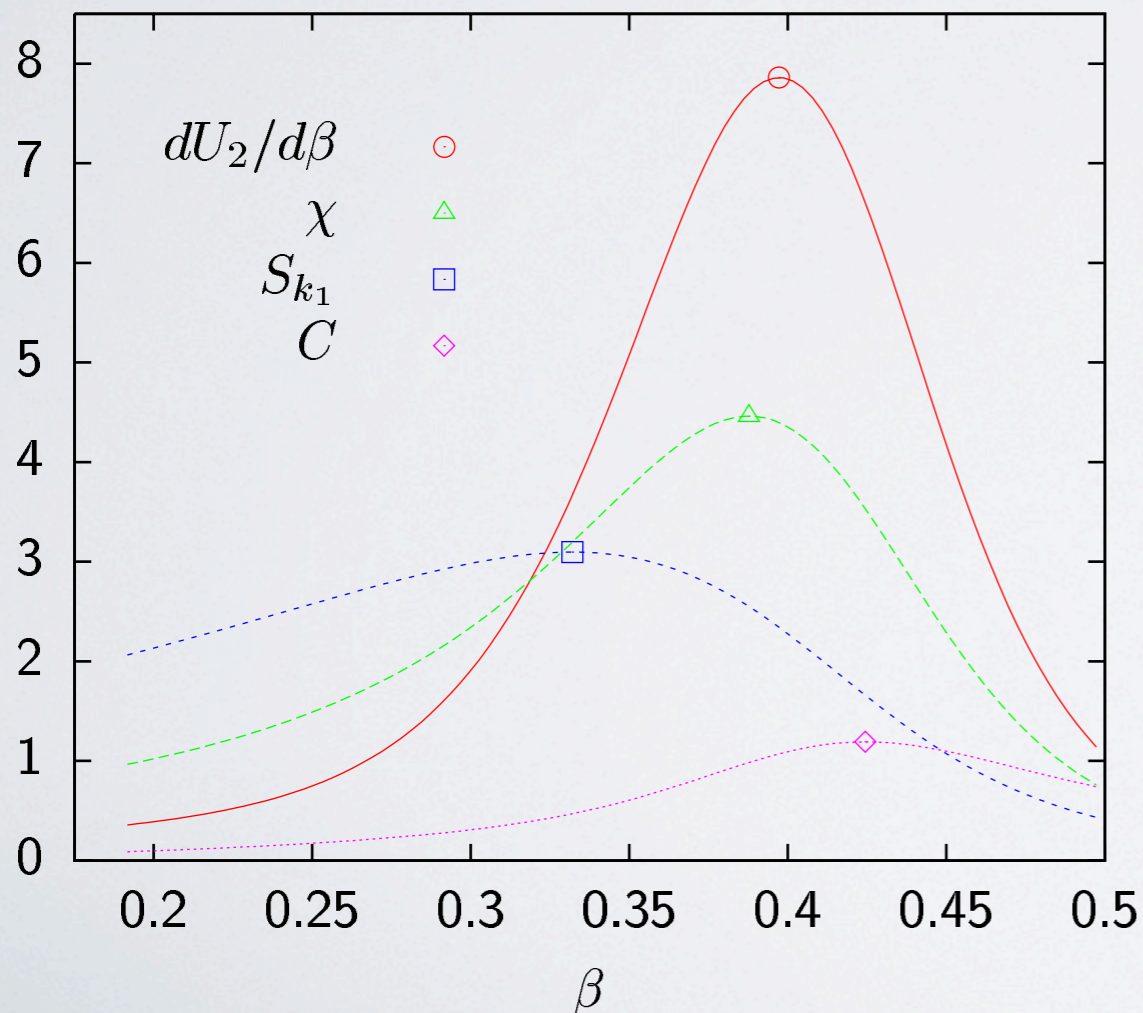
$$\chi(\beta) = \beta V (\langle m^2 \rangle - \langle |m| \rangle^2)$$

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

...

# Temperature Interval

- ❖ cover the complete desired “critical” temperature range



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

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

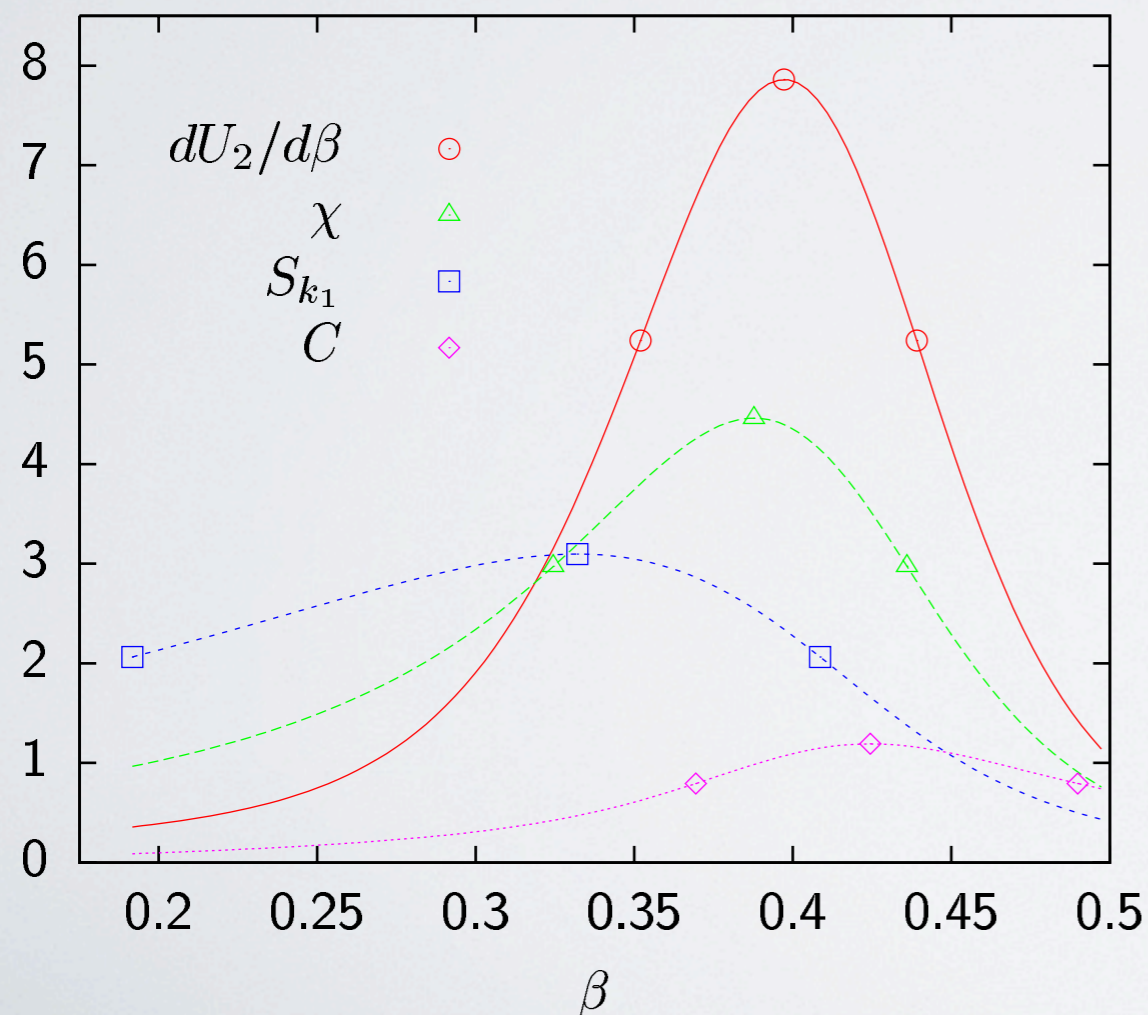
$$S(\beta_S^{+/-}) = r S^{\max}$$

$$\beta_S^+ > \beta_S^{\max} \quad \text{and} \quad \beta_S^- < \beta_S^{\max}$$

$$\text{“desired” range: } [\beta_{S_{k_1}}^-, \beta_C^+]$$

# Temperature Interval

- ❖ cover the complete desired “critical” temperature range



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

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

$$r = \frac{2}{3}$$

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

$$\beta_S^+ > \beta_S^{\max} \quad \text{and} \quad \beta_S^- < \beta_S^{\max}$$

“desired” range:  $[\beta_{S_{k_1}}^-, \beta_C^+]$

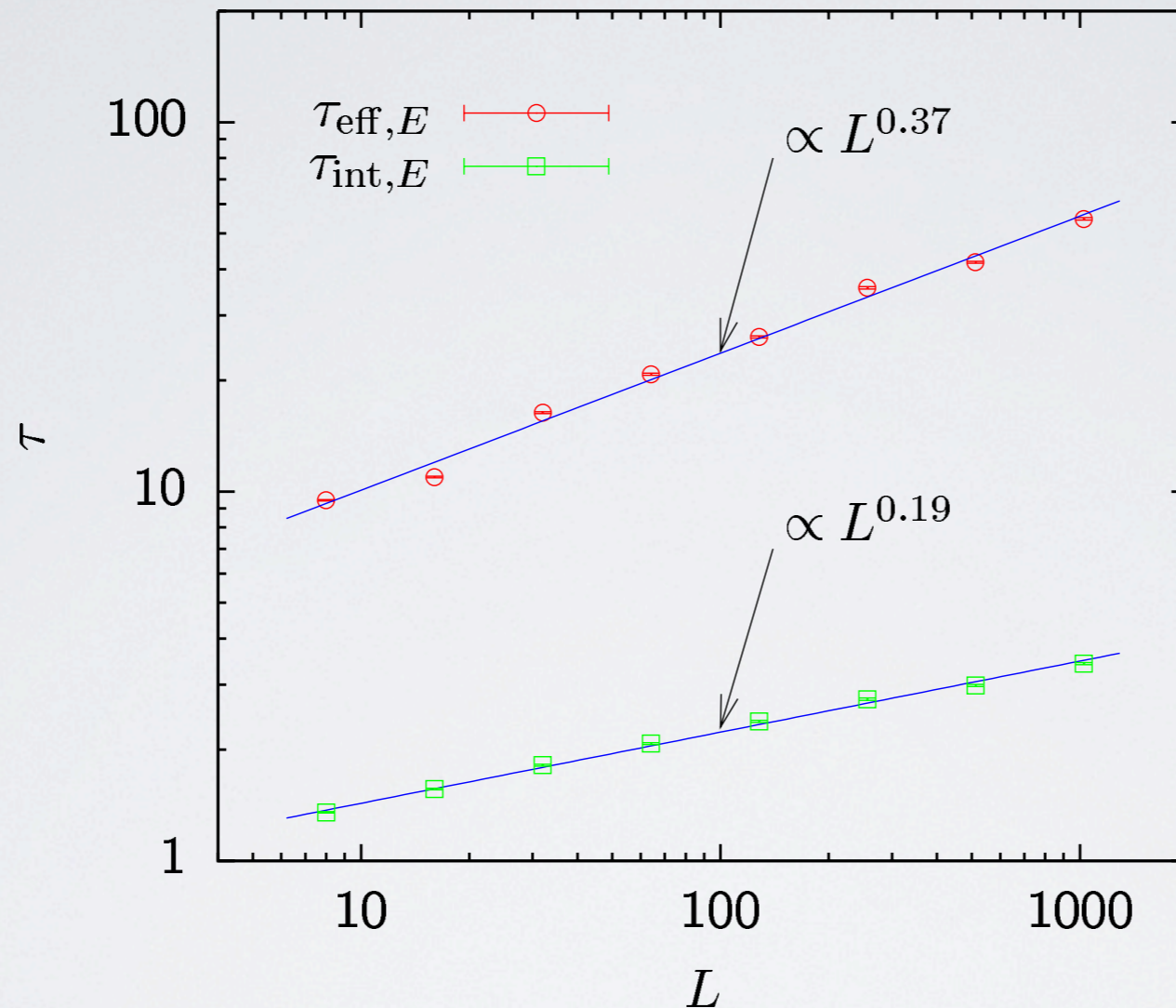


# Temperature Interval

general recipe:

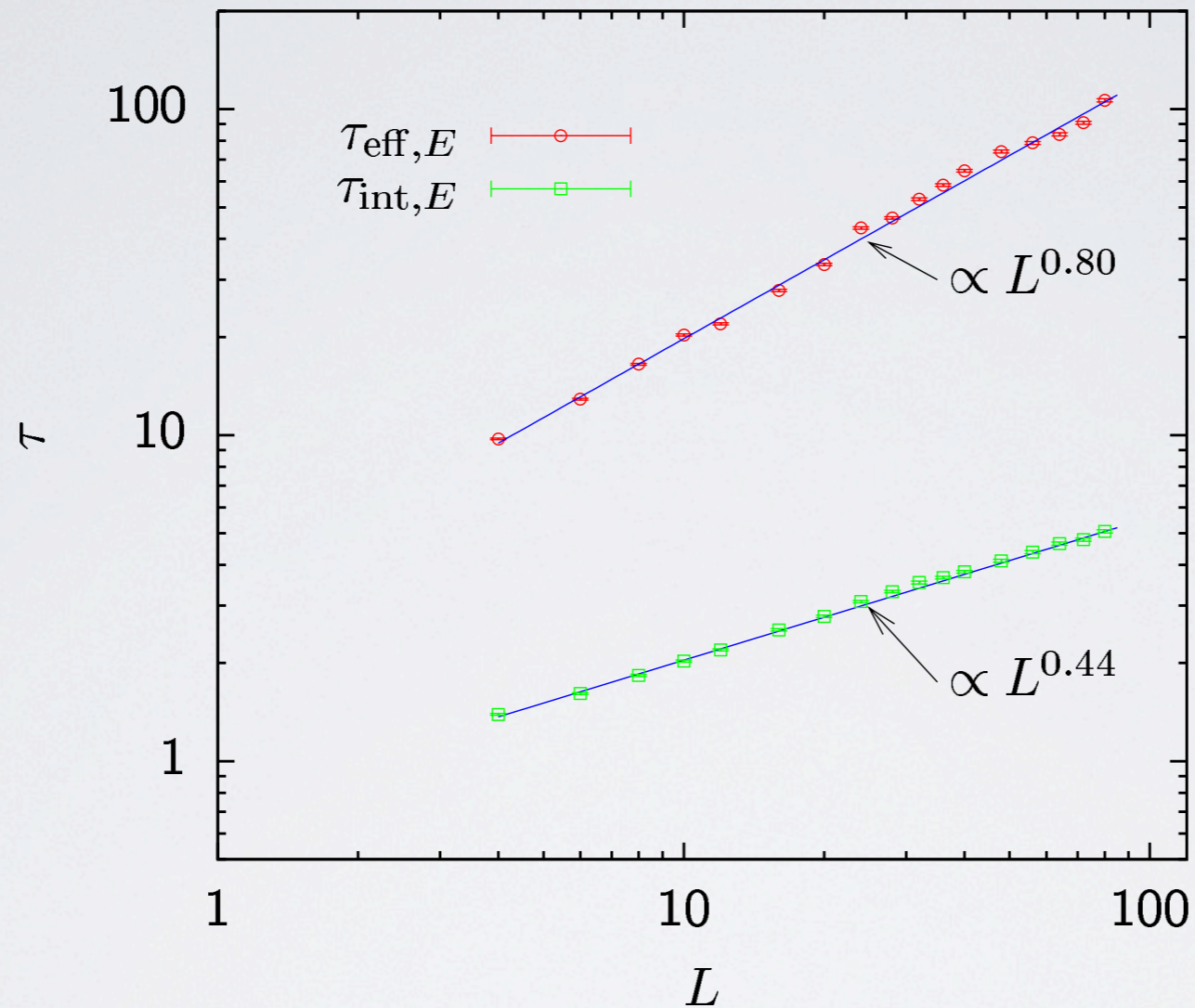
1. compute the simulation temperatures of the replica equidistant in  $\beta$ ,
2. perform several hundred thermalization sweeps and a short measurement run,
3. check the histogram overlap between adjacent replica: if the overlap is too small, add on or two replica and goto step 1, else go on,
4. use multi-histogram reweighting to determine  $\beta_S^-$  and  $\beta_S^+$  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  $\beta_i$  with fixed acceptance rate  $A(1 \rightarrow 2)$  until  $\beta_i = \beta^+ \geq \beta_{\max}^+$ ,
7. perform several hundred thermalization sweeps and a long measurement run.

# Autocorrelation times



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

# Autocorrelation times



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# Summary

What can we do to improve the parallel tempering algorithm?

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

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

EB and W. Janke, in preparation

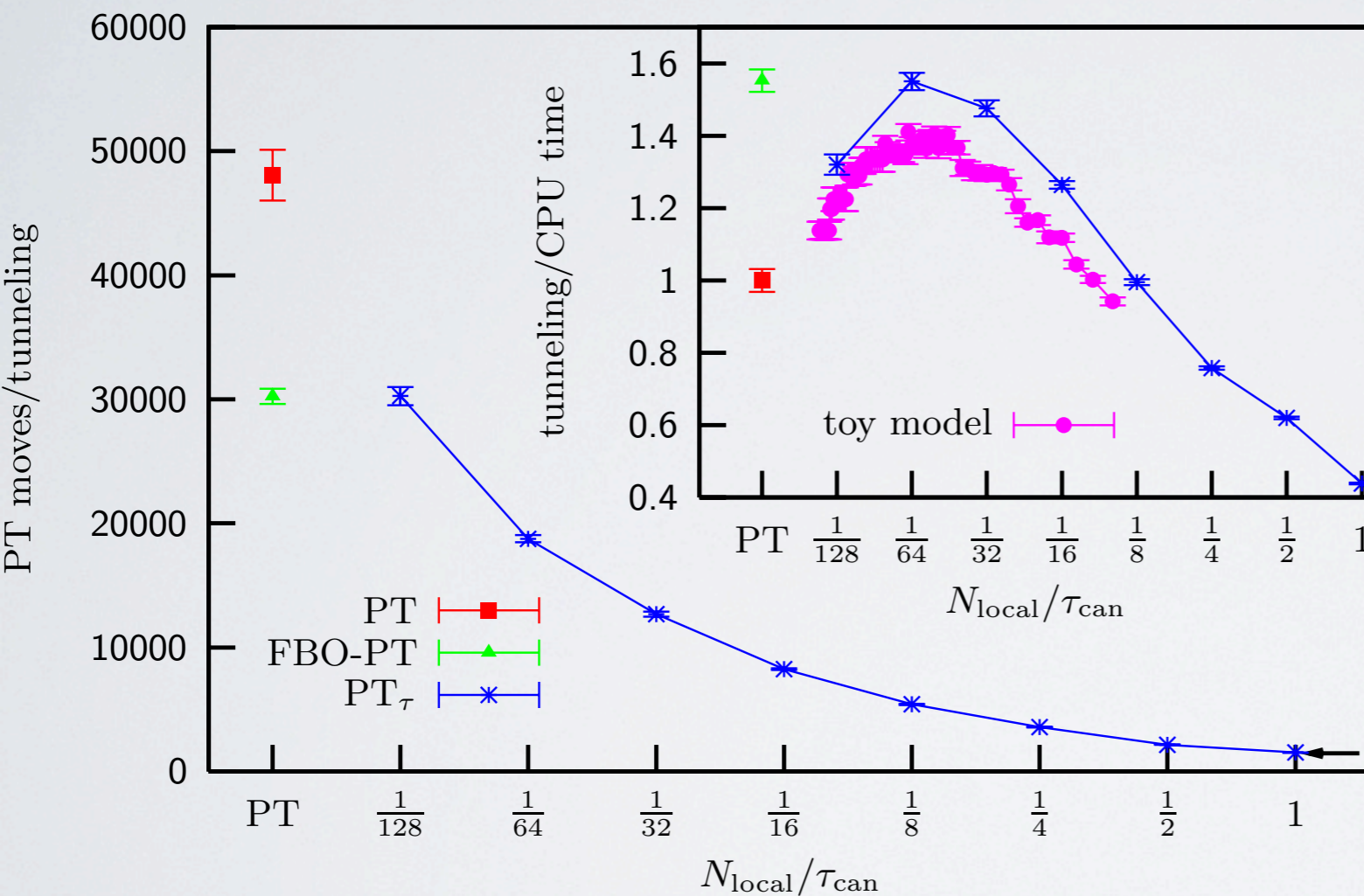
# Acknowledgements

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- EU-RTN Network “ENRAGE”

THANK YOU!

# Sweeps per tunneling



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