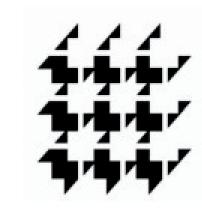


## Coupling from the Past for Statistical Mechanics Models

DIMACS Center for Discrete Mathematics and Theoretical Computer Science Founded as a National Science Foundation Science and Technology Center

Jasmine Khalil, Dr. Pierre Bellec

Pennsylvania State University, Rutgers University

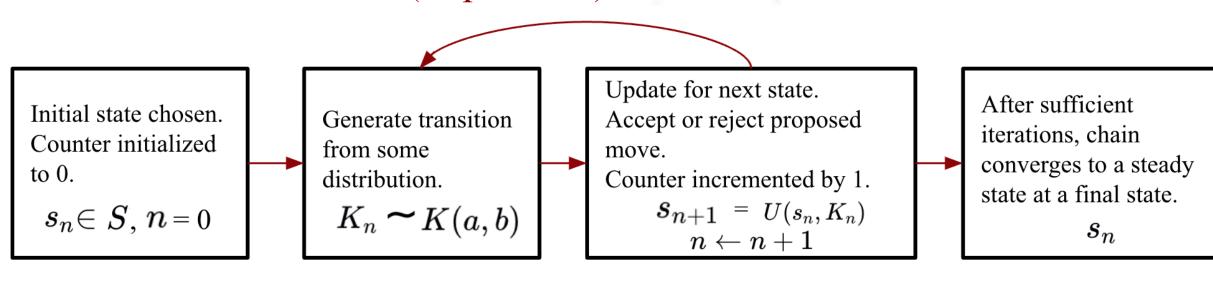


## Markov Chain Monte Carlo (MCMC)

Markov Chain - Process of stochastic transitions from one state to another. - Each transition is **memoryless**.

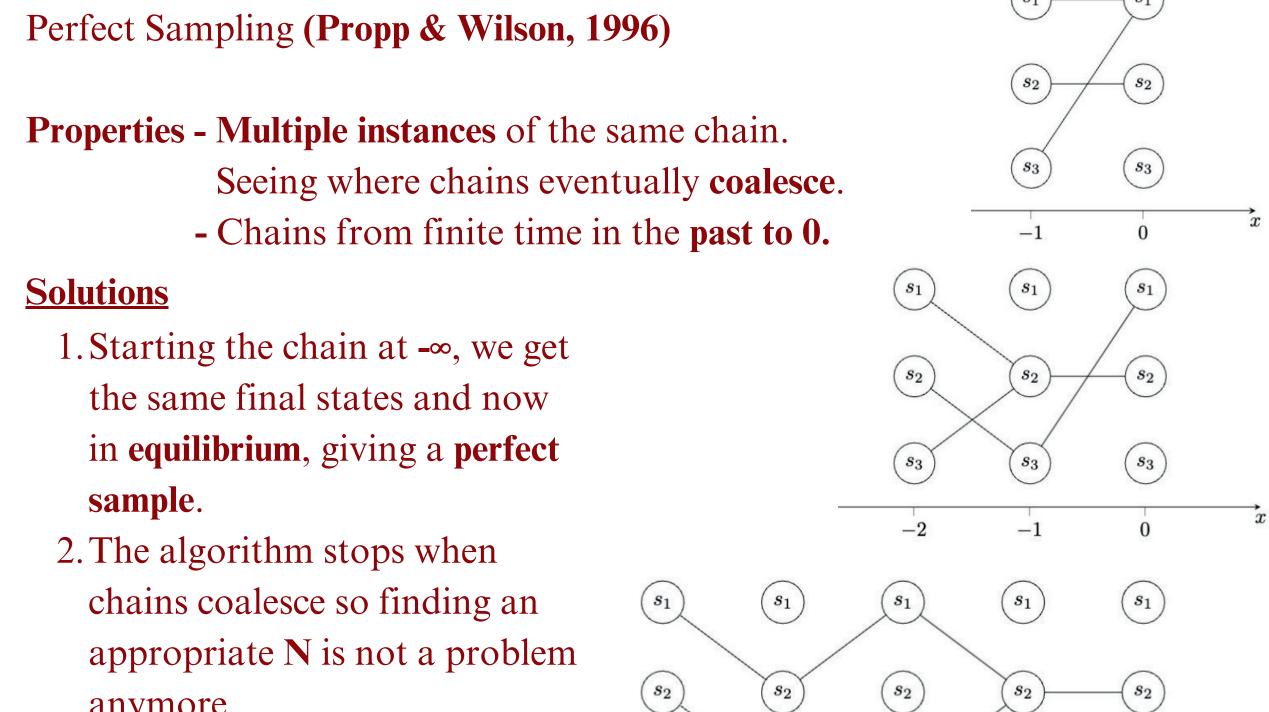
Distribution discrepancies: Not sampling from the exact distribution desired.  $\tilde{\pi}^{(n)} \neq \pi$ 

> 2. Upper bound on transition steps: Difficult to find the upper bound of N (steps taken).  $|\tilde{\pi}^{(n)} - \pi| < \epsilon$



#### Typical MCMC Algorithm Process.

## Coupling From the Past (CFTP)



# anymore. Steps Further in the Past till Coupling.

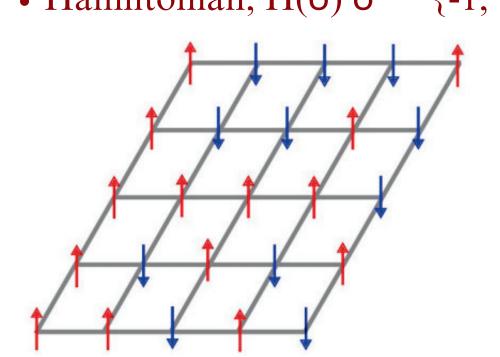
## The Ising Model

Let G = (V, E) be a graph.

Randomly assign -1's and +1's to the vertices of G representing dipoles in a **ferromagnetic material**.

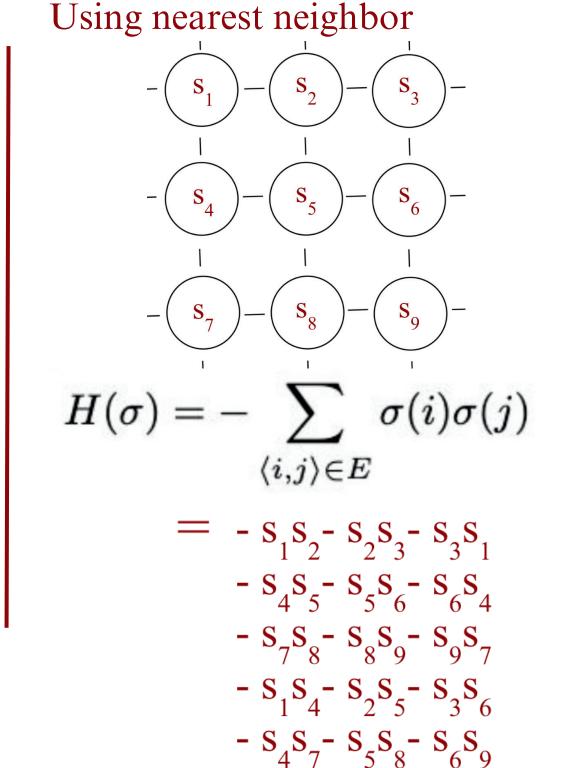
Probability distributions depend on:

- Inverse temperature,  $\beta \ge 0$
- Hamiltonian,  $H(\sigma)$   $\sigma$ **{-1, 1}**



Lattice of magnetic dipole moments, each either spin up (+1) or down (-1).

**Energy (Hamiltonian)** 



 $-S_{7}S_{1}-S_{8}S_{2}-S_{9}S_{3}$ 

**Update Function** Accept or reject a

spin (state) change.

- Let X'(v) = s and  $X'(w) = X_t(w), w \neq v$ .
- Set  $X_{t+1} = \begin{cases} X' & \text{if } r \leq \min\{1, e^{-\beta H(X')} / e^{-\beta H(X_t)}\} \\ X_t & \text{otherwise} \end{cases}$

## The Ising Model

#### **Phase Transition Phenomena**

Low temperatures (high  $\beta$ ), spontaneous magnetization.

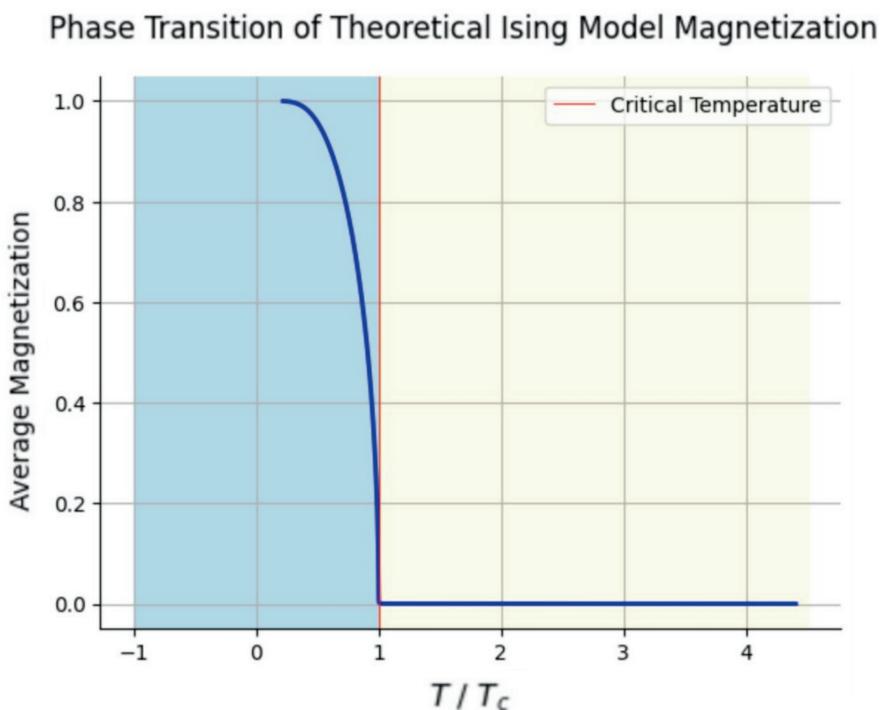
High temperatures (low  $\beta$ ), magnetization is entirely lost.

#### **Onsager Critical Value:**

$$\beta_c = \frac{1}{2}log(1+\sqrt{2}) \approx 0.441$$

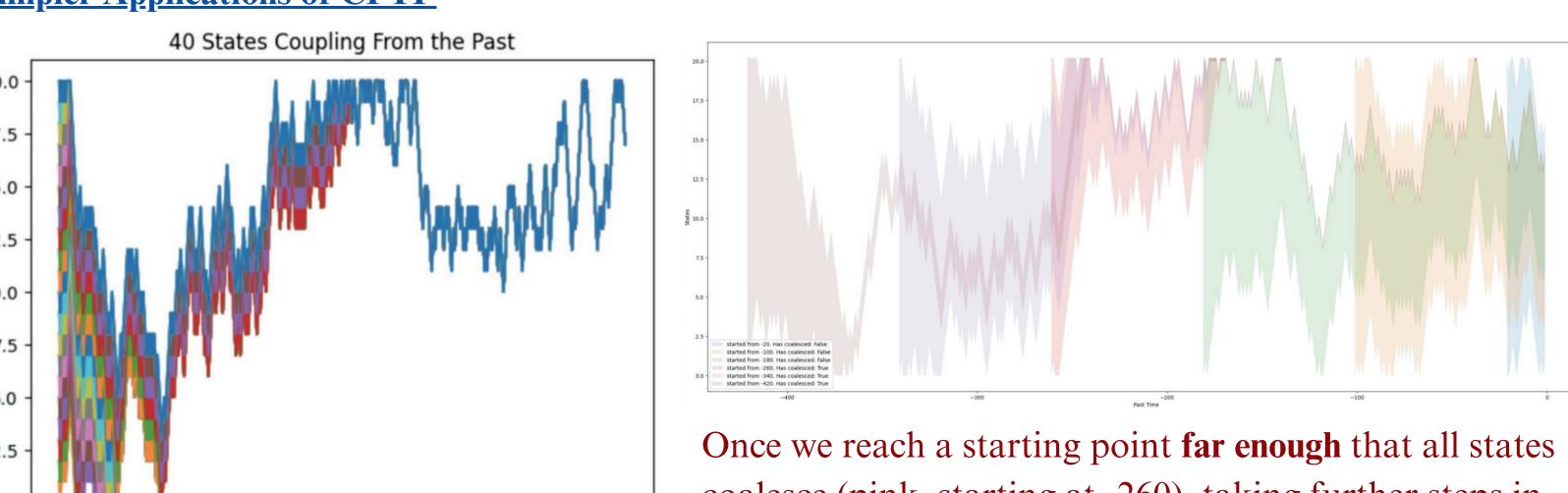
**Theoretical Average Magnetization Iteration** Scheme:

$$s_{i+1} = anh \left\{ rac{T_c}{T} (rac{H}{H_c} + s_i) 
ight\}$$
 $s_{i+1} = anh \left\{ rac{T_c}{T} s_i 
ight\} igotarrow ext{Neglecting}$ 
External
Magnetic Field



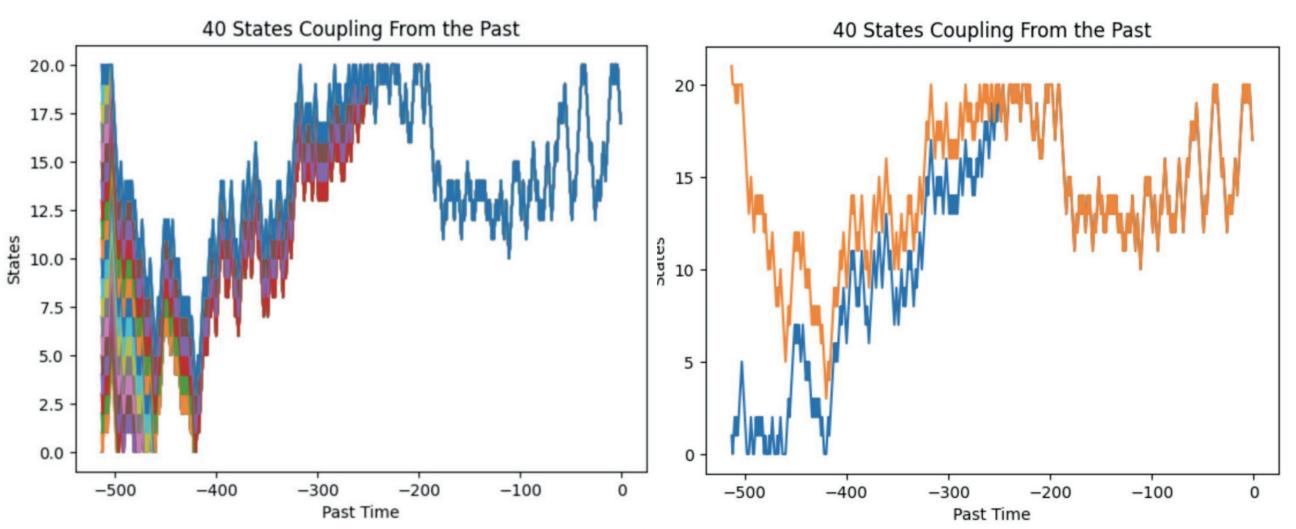
## Implementation - Simple Applications

#### **Simpler Applications of CFTP**



**Monotone CFTP - Sandwiching** 

- coalesce (pink, starting at -260), taking further steps in the past produces the same result (violet and brown starting at -340 and -420, respectively).
- Running k separate instances of a chain becomes **practically difficult** as k reaches **large values**.
- For Markov chains obeying monotonicity properties, sandwiching applies CFTP focusing only on the two extreme states.



• Choose  $v \in_R V$ ,  $s \in_R \{+1, -1\}$  and  $r \in_R [0, 1]$ .

Vigoda 2003

Two Extreme States

For a 3x3 lattice: 512

suggests that the two

bounding states are:

all states spin up

spin down (-1).

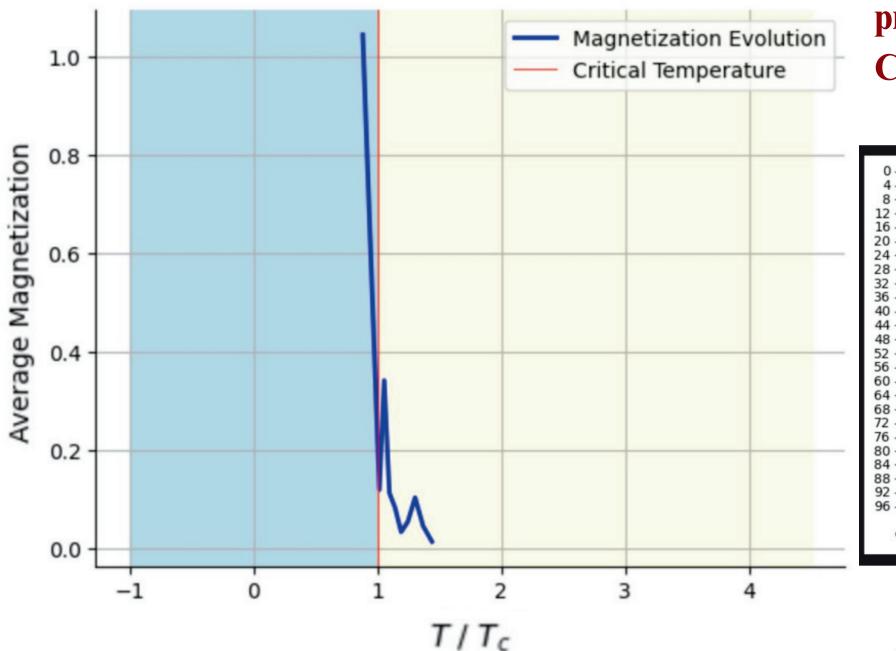
(+1), and all states

possible states.

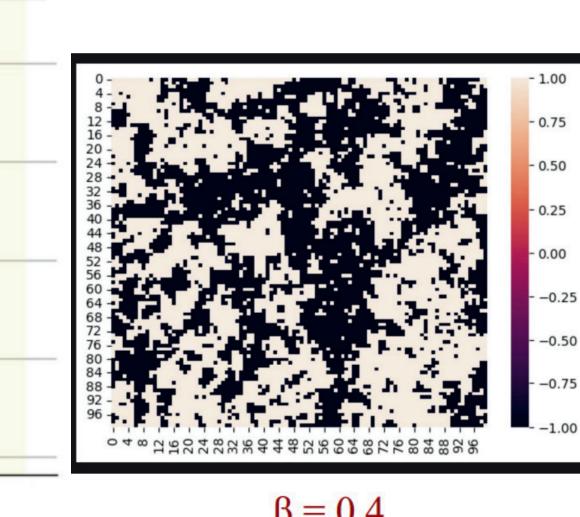
Monotonicity

## Implementation - The Ising Model

Phase Transition of Practical Ising Model Magnetization



Phase transition phenomena of practical ising model using the **CFTP** algorithm.



Our simulation of CFTP for a  $100 \times 100$  lattice at  $\beta = 0.4$ , close to the critical temperature

#### **Update Test Simplification:**

- Unsupported Python features.

Using Numba before each time-

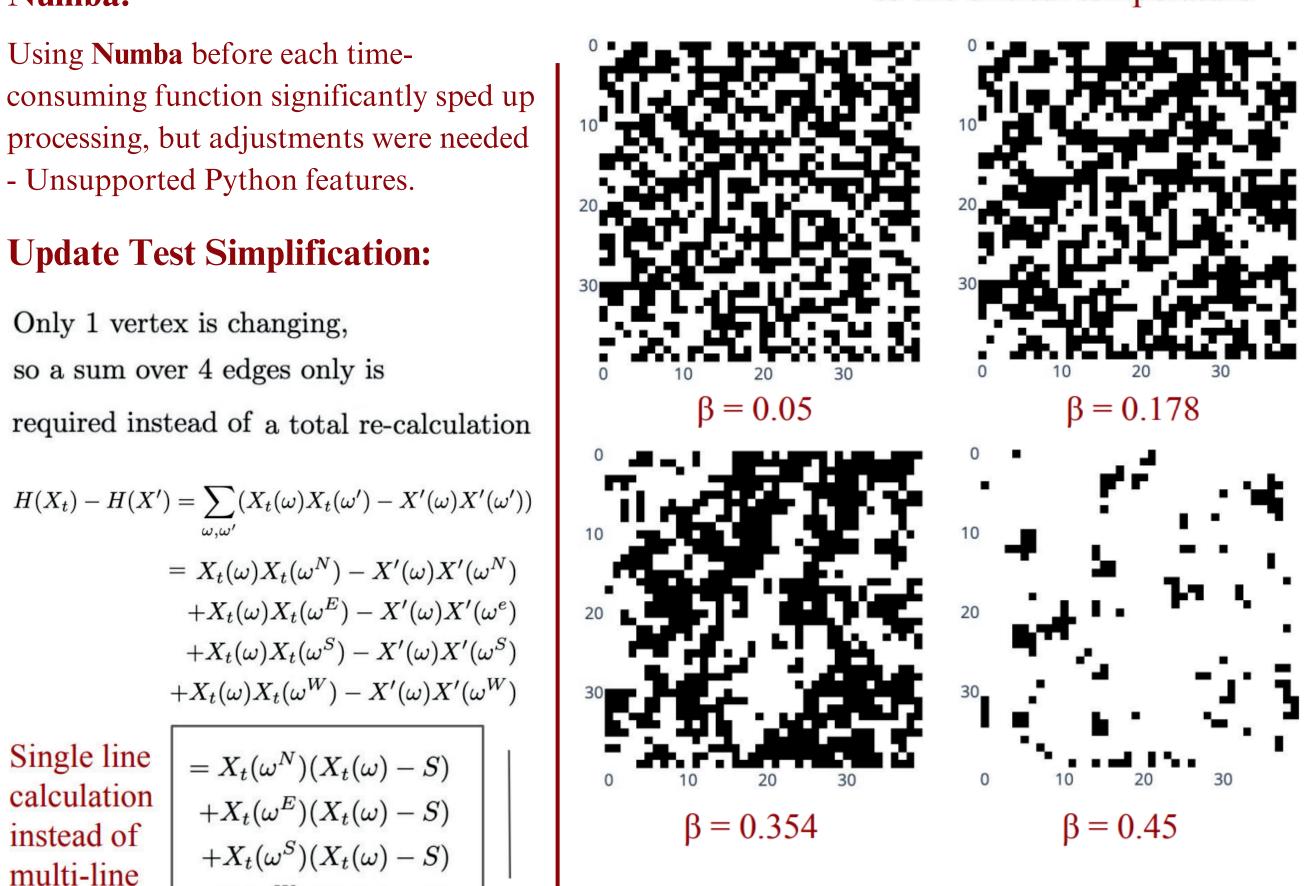
**Optimization** 

Numba:

Only 1 vertex is changing, so a sum over 4 edges only is required instead of a total re-calculation

 $H(X_t) - H(X') = \sum_{t} (X_t(\omega)X_t(\omega') - X'(\omega)X'(\omega'))$  $= X_t(\omega)X_t(\omega^N) - X'(\omega)X'(\omega^N)$  $+X_t(\omega)X_t(\omega^E)-X'(\omega)X'(\omega^e)$  $+X_t(\omega)X_t(\omega^S)-X'(\omega)X'(\omega^S)$  $+X_t(\omega)X_t(\omega^W)-X'(\omega)X'(\omega^W)$ 

 $=X_t(\omega^N)(X_t(\omega)-S)$ calculation  $+X_t(\omega^E)(X_t(\omega)-S)$ instead of  $+X_t(\omega^S)(X_t(\omega)-S)$ multi-line  $+X_t(\omega^W)(X_t(\omega)-S)$ iterations



Progression of 40x40 lattice at different  $\beta$ 

### Acknowledgments & References

This work was carried out as a part of the 2024 DIMACS REU program at Rutgers University, supported by NSF grant CNS-2150186.

Thank you to my REU advisor, **Dr. Pierre C. Bellec**, for his guidance!

Cipra, Barry A. 1987. "An introduction to the Ising Model." The American Mathematical Monthly, Vol. 94, No. 10, pp. 937-959. Cordaro, Dylan. 2017. "Markov Chains and Coupling from the Past."

Dembo, A.; Funaki, T.; Picard, J. 2005. "Lectures on probability theory and statistics" Ecole d'eté De Probabilités de Saint-Flour XXXIII-2003; Springer: Berlin.

Häggström, Olle. 2002. "Finite Markov Chains and Algorithmic Applications." Vol. 52. Cambridge University Press. Levin, D. A.; Peres, Y.; Wilmer, E. L.; Andersson, S. 2018. "Markov chains and mixing times." MTM: Johanneshov 348-58. Propp, James Gary, and David Bruce Wilson. 1996. "Exact Sampling with Coupled Markov Chains and Applications to Statistical Mechanics." Random Structures & Algorithms 9 (1-2): 223–52.

Propp, James, and David Wilson. 1997. "Coupling from the Past: A User's Guide." Microsurveys in Discrete Probability 41:

Puttick, Alexandre R. 2009. "The Ising Model: Phase Transition in a Square Lattice." University of Chicago Schmidt, Volker. "Markov Chains and Monte-Carlo Simulation." Lecture Notes. Ulm University Institute of Stochastics. Summer

Stanley, H. Eugene. 1987. "Introduction to Phase Transitions and Critical Phenomena." Oxford University Press. Vigoda, Eric. "Coupling from the Past." CS37101-1 Markov Chain Monte Carlo Methods. Lecture 3. University of Chicago, October 14 2003.