

An approximation for binary Markov random fields

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Introduction

- Binary (hidden) Markov random field

$$\pi(x|\theta) = \frac{1}{b(\theta)} \exp \left\{ - \sum_{C \in \mathcal{C}} V_C(x_C|\theta) \right\}$$

- ex: Ising model (with external field)
- Likelihood evaluation not possible (MLE via MCMCMLE)
- Direct simulation not possible (simulation possible by MCMC)
- Our goals:
 - define approximate normalising constant
 - define approximate model from which direct simulation is possible
- Starting point: forward-backward recursion for MRF

Plan

- Notation
- Exact forward-backward recursion
- Approximate forward-backward recursion
- Evaluation criteria for approximation
 - compare to exact results (for small lattices)
 - compare results for different approximation levels
 - use approximation as proposal in independent proposal Metropolis–Hastings algorithm
- Simulation examples.

Notation

● Binary Markov random field on $n_I \times n_J$ lattice, $N = n_I \cdot n_J$

● State vector $x = (x_1, \dots, x_N)$, with $x_i \in \{0, 1\}$

● Set of cliques: \mathcal{C}

● Joint distribution

1	2	3	4	5
6	7	8	9	10
11	12	13		
			N-1	N

$$\pi(x|\theta) = \frac{a(x|\theta)}{b(\theta)} \propto a(x|\theta) = \exp \left\{ - \sum_{C \in \mathcal{C}} V_C(x_C, \theta) \right\} = \prod_{C \in \mathcal{C}} e^{-V_C(x_C, \theta)}$$

● Reformulation

$$a(x|\theta) = \prod_{k=1}^N a_k(x_{k:k+p+1}, \theta)$$

● p is a function of n_J and the clique size

● for first-order pairwise interaction model: $p = n_J$

Exact forward recursion

- Following Pettitt and Reeves (2004) and Friel and Rue (2007)
- For illustration: Small 6×5 lattice, Ising model

$$\pi(x_{1:30}|\theta) = b(\theta)^{-1} a_1(x_{1:6}, \theta) a_2(x_{2:7}, \theta) a_3(x_{3:8}, \theta) \cdots$$

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$$\pi(x_{2:30}|\theta) = b(\theta)^{-1} \left[\sum_{x_1} a_1(x_{1:6}, \theta) \right] a_2(x_{2:7}, \theta) a_3(x_{3:8}, \theta) \cdots$$

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$$\pi(x_{3:30}|\theta) = b(\theta)^{-1} \left[\sum_{x_2} b_2(x_{2:6}, \theta) a_2(x_{2:7}, \theta) \right] a_3(x_{3:8}, \theta) \cdots$$

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$$\pi(x_{3:30}|\theta) = b(\theta)^{-1} b_3(x_{3:6}, \theta) a_3(x_{3:8}, \theta) \cdots$$

⋮

$$\pi(x_{30}|\theta) = b(\theta)^{-1} b_{30}(x_{30}, \theta) a_{30}(x_{30}, \theta)$$

Exact backward recursion

- Following Pettitt and Reeves (2004) and Friel and Rue (2007)
- For illustration: Small 6×5 lattice, Ising model

$$\pi(x_{30}|\theta) = b(\theta)^{-1} b_{30}(x_{30}, \theta) a_{30}(x_{30}, \theta) \propto b_{30}(x_{30}, \theta) a_{30}(x_{30}, \theta)$$

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$$\begin{aligned}\pi(x_{29}|x_{30}, \theta) &\propto \pi(x_{29:30}|\theta) = b_{29}(x_{29:30}, \theta)a_{29}(x_{29:30}, \theta)a_{30}(x_{30}, \theta) \\ &\propto b_{29}(x_{29:30}, \theta)a_{29}(x_{29:30}, \theta)\end{aligned}$$

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$$\pi(x_{28}|x_{29:30}, \theta) \propto b_{28}(x_{28:30}, \theta)a_{28}(x_{28:30}, \theta)$$

Exact backward recursion

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$$\propto b_{29}(x_{29:30}, \theta)a_{29}(x_{29:30}, \theta)$$

$$\pi(x_{28}|x_{29:30}, \theta) \propto b_{28}(x_{28:30}, \theta)a_{28}(x_{28:30}, \theta)$$

⋮

$$\pi(x_2|x_{2:30}, \theta) \propto b_2(x_{2:6}, \theta)a_2(x_{2:7}, \theta)$$

$$\pi(x_1|x_{2:30}, \theta) \propto b_1(x_{1:5}, \theta)a_1(x_{1:6}, \theta)$$

Exact backward recursion

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⋮

$$\pi(x_2|x_{2:30}, \theta) \propto b_2(x_{2:6}, \theta) a_2(x_{2:7}, \theta)$$

$$\pi(x_1|x_{2:30}, \theta) \propto b_1(x_{1:5}, \theta) a_1(x_{1:6}, \theta)$$

- Computational complexity: forward: $O(N2^p)$, backward: cheap

Approximate forward recursion

- For illustration: Small 6×5 lattice, Ising model

$$\pi(x_{1:30}|\theta) = b(\theta)^{-1} a_1(x_{1:6}, \theta) a_2(x_{2:7}, \theta) a_3(x_{3:8}, \theta) \cdots$$

$$\begin{aligned} \pi(x_{2:30}|\theta) &= b(\theta)^{-1} \left[\sum_{x_1} a_1(x_{1:6}, \theta) \right] a_2(x_{2:7}, \theta) a_3(x_{3:8}, \theta) \cdots \\ &= b(\theta)^{-1} b_2(x_{2:6}, \theta) a_2(x_{2:7}, \theta) a_3(x_{3:8}, \theta) \cdots \end{aligned}$$

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⋮

$$\pi(x_{30}|\theta) \approx \tilde{b}(\theta)^{-1} \tilde{b}_{30}(x_{30}, \theta) a_{30}(x_{30}, \theta)$$

Approximate backward recursion

- For illustration: Small 6×5 lattice, Ising model

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⋮

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$$\pi(x_1|x_{2:30}, \theta) \propto \tilde{b}_1(x_{1:5}, \theta)a_1(x_{1:6}, \theta)$$

Representation of $b_k(x_{k:k+p}, \theta)$

- Exact recursive formula

$$b_{k+1}(x_{k+1:k+p+1}, \theta) = \sum_{x_k} b_k(x_{k:k+p}, \theta) a_k(x_{k:k+p+1}, \theta)$$

- Number of values to store for each k : 2^p
- Represent $b_k(x_{k:k+p}, \theta)$ as

$$\begin{aligned} \ln \{b_k(x_{k:k+p}, \theta)\} &= \beta_k^\emptyset + \sum_{i=k}^{k+p} \beta_k^{\{i\}} x_i + \sum_{i=k}^{k+p-1} \sum_{j=i+1}^{k+p} \beta_k^{\{i,j\}} x_i x_j \\ &+ \sum_{i=k}^{k+p-2} \sum_{j=i+1}^{k+p-1} \sum_{t=j+1}^{k+p} \beta_k^{\{i,j,t\}} x_i x_j x_t + \dots + \beta_k^{\{k,\dots,k+p\}} x_k \cdot \dots \cdot x_{k+p} \\ &= \sum_{\Lambda \subseteq \{k,\dots,k+p\}} \beta_k^\Lambda \prod_{i \in \Lambda} x_i \end{aligned}$$

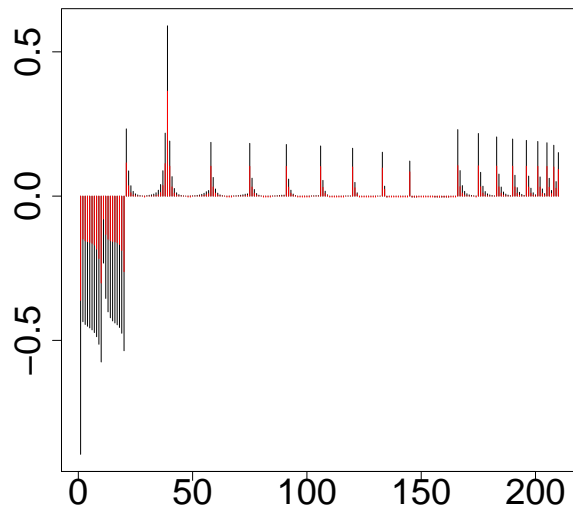
Ising example

● Ising model — small 20×20 lattice, i.e. $p = 20$

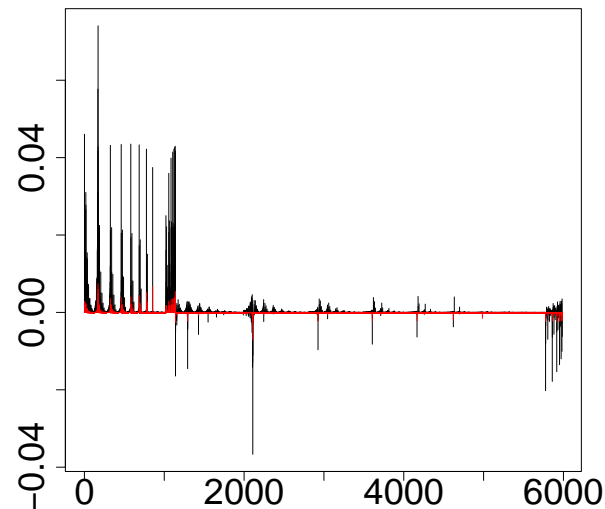
● red: $\beta = 0.6$

● black: $\beta = 0.8$

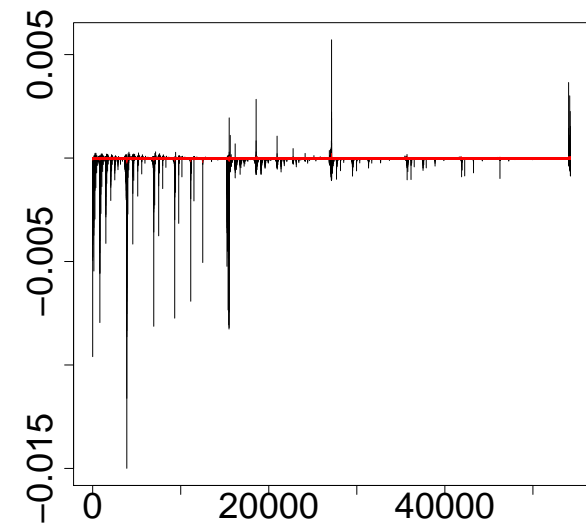
1. and 2.



3. and 4.



5. and 6.



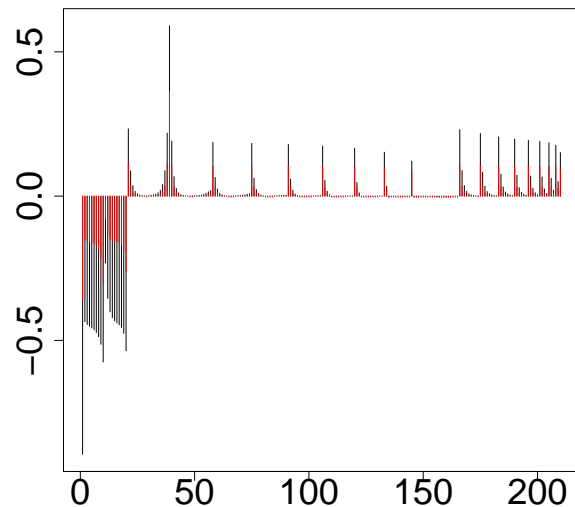
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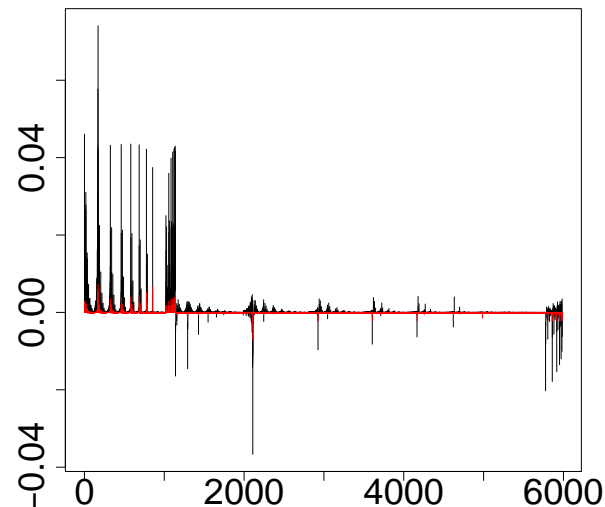
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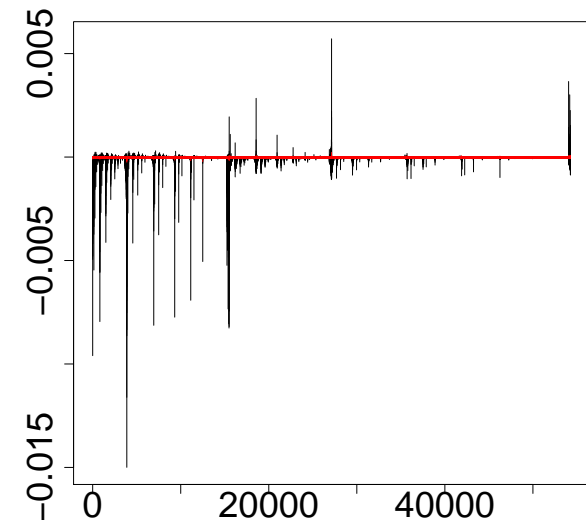
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● $\beta = 0.8$: fraction of interactions larger than ε

ε	10^{-1}	10^{-2}	10^{-3}	10^{-4}	10^{-5}
fraction	0.000039	0.000169	0.000906	0.00556	0.0314

Forward recursion for new representation

- Correspondingly we represent

$$\ln \{a_k(x_{k:k+p+1}, \theta)\} = \sum_{\Lambda \subseteq \{k, \dots, k+p+1\}} \alpha_k^\Lambda \prod_{i \in \Lambda} x_i$$

- Old recursion formula

$$b_{k+1}(x_{k+1:k+p+1}, \theta) = \sum_{x_k} b_k(x_{k:k+p}, \theta) a_k(x_{k:k+p+1}, \theta)$$

- Recursion formula for new representation

$$\begin{aligned} & \exp \left\{ \sum_{\Lambda \subseteq \{k+1, \dots, k+p+1\}} \beta_{k+1}^\Lambda \prod_{i \in \Lambda} x_i \right\} \\ &= \sum_{x_k} \exp \left\{ \sum_{\Lambda \subseteq \{k, \dots, k+p\}} \beta_k^\Lambda \prod_{i \in \Lambda} x_i + \sum_{\Lambda \subseteq \{k, \dots, k+p+1\}} \alpha_k^\Lambda \prod_{i \in \Lambda} x_i \right\} \end{aligned}$$

Recursive computation of β_{k+1}^Λ

● Setting $x_{k+1} = \dots = x_{k+p+1} = 0$ gives

$$\exp \left\{ \beta_{k+1}^\emptyset \right\} = \exp \left\{ \beta_k^\emptyset + \alpha_k^\emptyset \right\} + \exp \left\{ \beta_k^\emptyset + \beta_k^{\{k\}} + \alpha_k^\emptyset + \alpha_k^{\{k\}} \right\}$$

Recursive computation of β_{k+1}^Λ

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- Setting $x_i = 1$ and $x_j = 0$ for all $j \neq i$ gives

$$\exp \left\{ \beta_{k+1}^\emptyset + \beta_{k+1}^{\{i\}} \right\} = \exp \left\{ \beta_k^\emptyset + \beta_k^{\{i\}} + \alpha_k^\emptyset + \alpha_k^{\{i\}} \right\}$$

$$+ \exp \left\{ \beta_k^\emptyset + \beta_k^{\{i\}} + \beta_k^{\{k\}} + \beta_k^{\{k,i\}} + \alpha_k^\emptyset + \alpha_k^{\{i\}} + \alpha_k^{\{k\}} + \alpha_k^{\{k,i\}} \right\}$$

Recursive computation of β_{k+1}^Λ

- Setting $x_{k+1} = \dots = x_{k+p+1} = 0$ gives

$$\exp \left\{ \beta_{k+1}^\emptyset \right\} = \exp \left\{ \beta_k^\emptyset + \alpha_k^\emptyset \right\} + \exp \left\{ \beta_k^\emptyset + \beta_k^{\{k\}} + \alpha_k^\emptyset + \alpha_k^{\{k\}} \right\}$$

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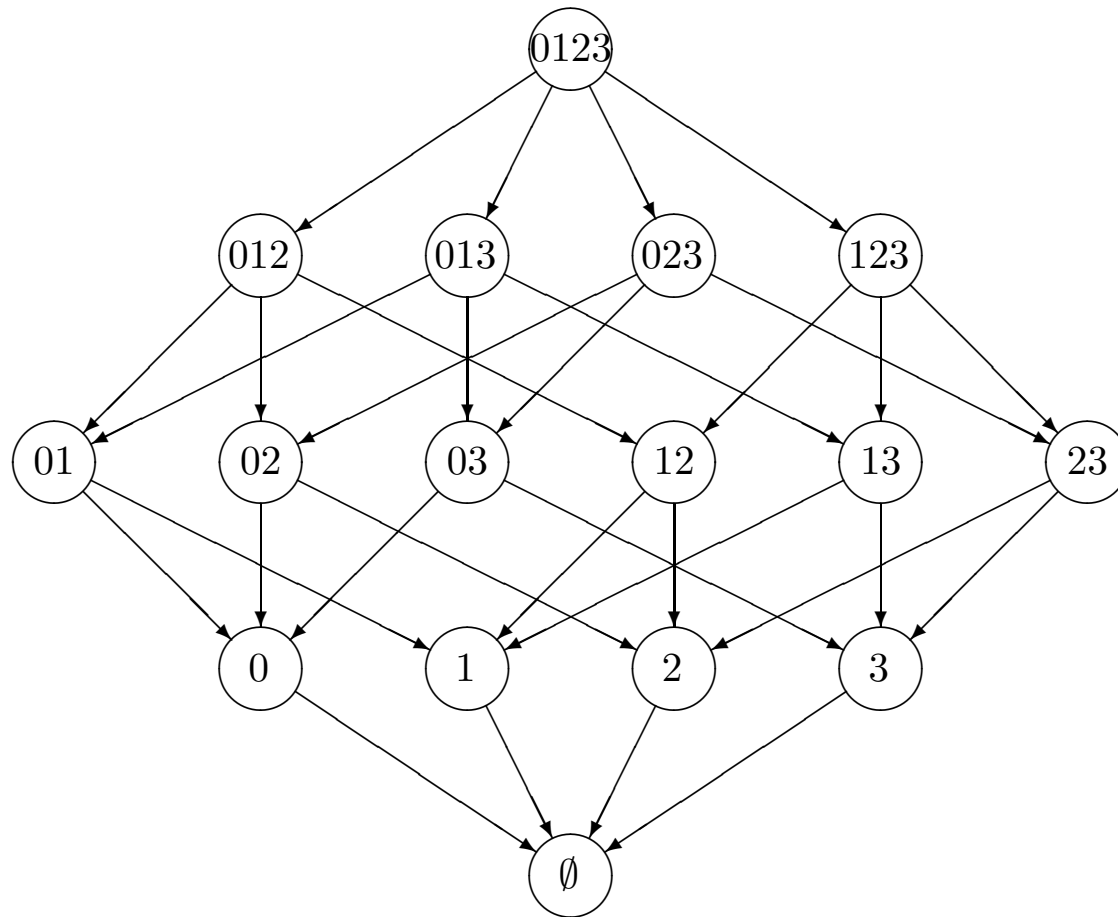
$$+ \exp \left\{ \beta_k^\emptyset + \beta_k^{\{i\}} + \beta_k^{\{k\}} + \beta_k^{\{k,i\}} + \alpha_k^\emptyset + \alpha_k^{\{i\}} + \alpha_k^{\{k\}} + \alpha_k^{\{k,i\}} \right\}$$

- Setting $x_i = x_j = 1$ and $x_t = 0$ for $t \notin \{i, j\}$ gives

$$\exp \left\{ \beta_{k+1}^\emptyset + \beta_{k+1}^{\{i\}} + \beta_{k+1}^{\{j\}} + \beta_{k+1}^{\{i,j\}} \right\} = \dots$$

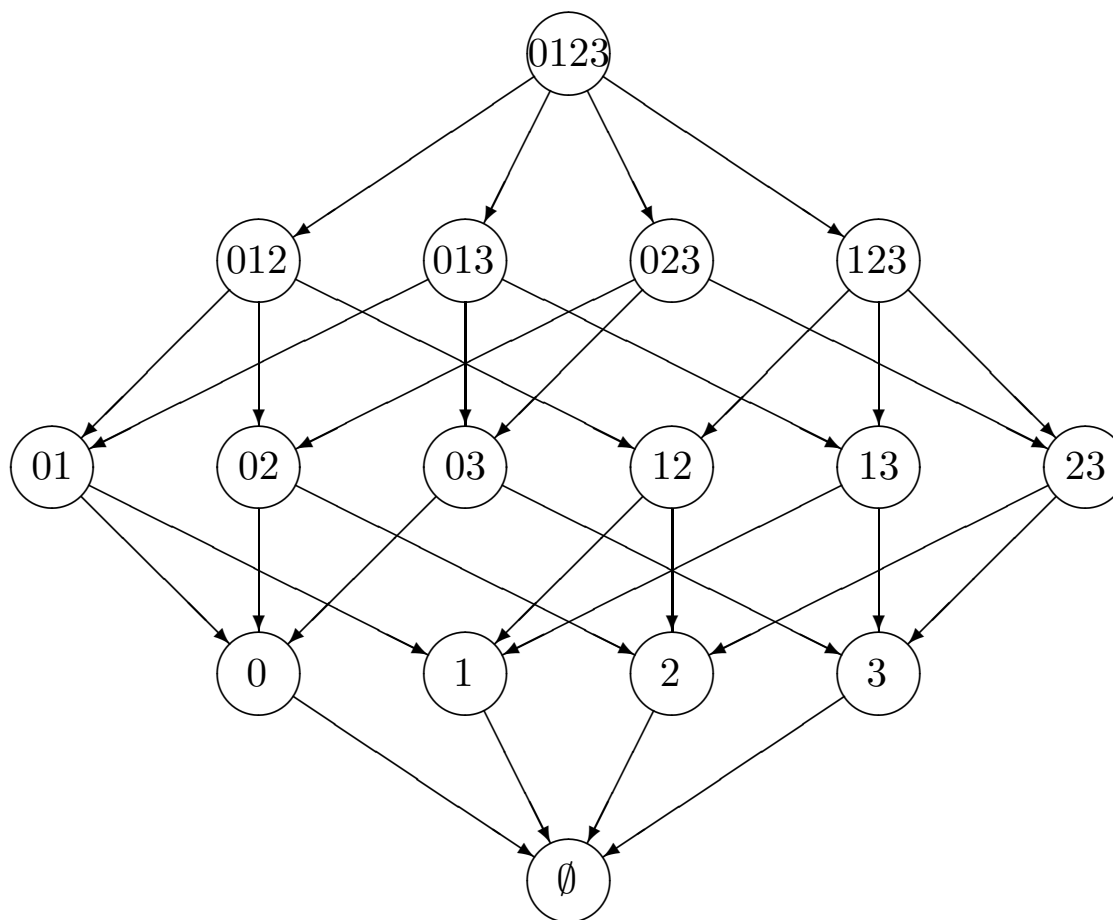
Recursive scheme

$$\ln \{b_k(x_{k:k+p}, \theta)\} = \sum_{\Lambda \subseteq \{k, \dots, k+p\}} \beta_k^\Lambda \prod_{i \in \Lambda} x_i$$



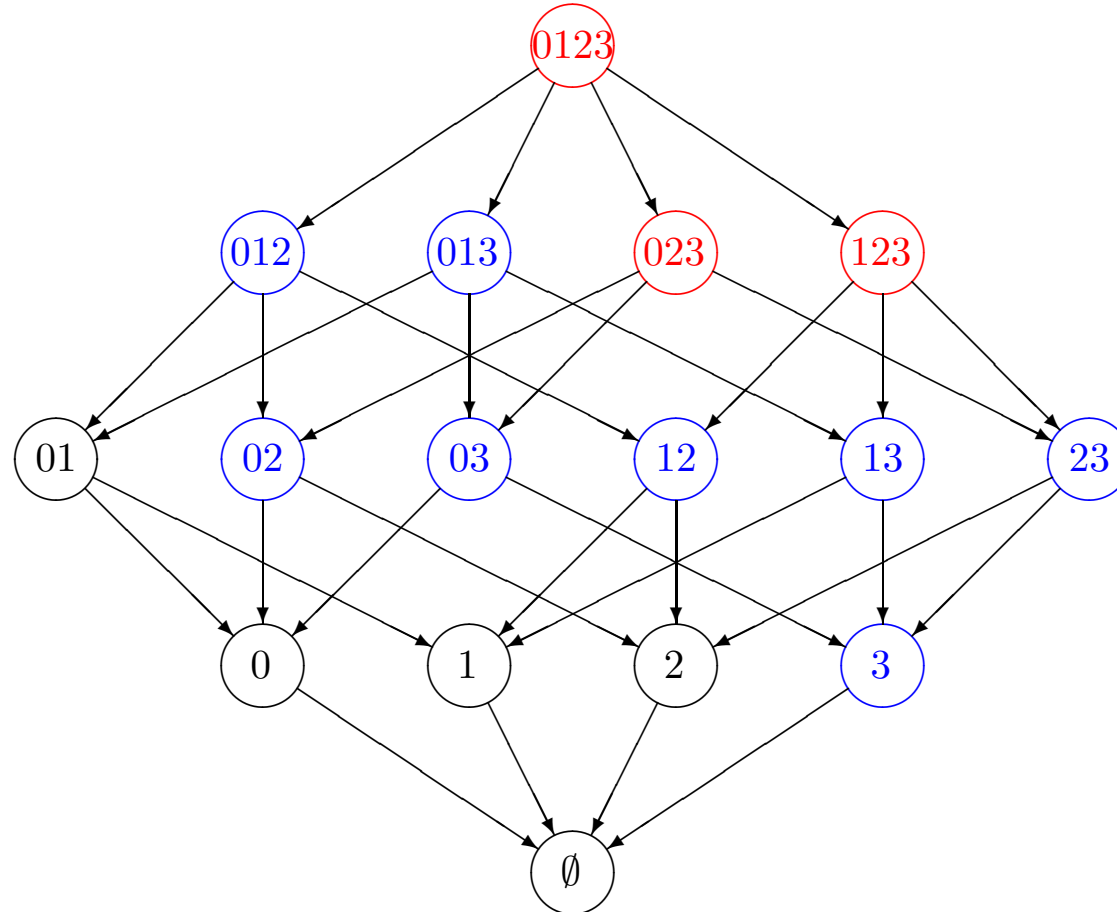
Approximation strategy

- If a computed β_k^Λ has $|\beta_k^\Lambda| \leq \varepsilon$, approximate β_k^Λ to zero
- If for some Λ , β_k^λ is (approximated to) zero for all $\lambda \subset \Lambda$, $|\lambda| = |\Lambda| - 1$, approximate β_k^Λ to zero (without computing the exact value)



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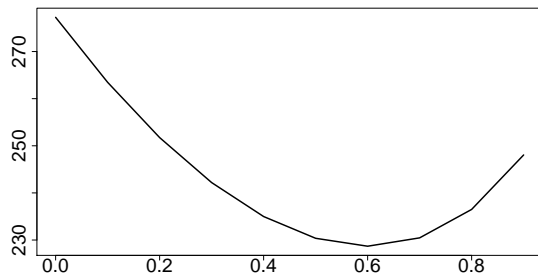
Evaluation criteria for approximation

- Compare approximation with exact result — for small lattices
- Compare results for different cut off values ε
- Use approximation as proposal in an independent proposal Metropolis–Hastings algorithm — look at acceptance rate

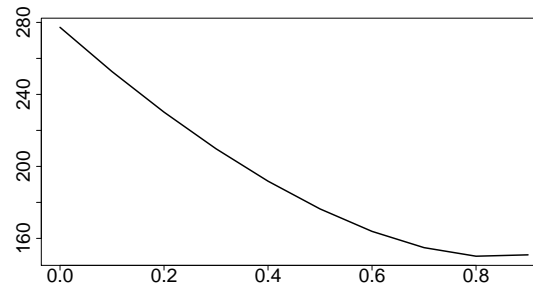
Comparison to exact results in a small lattice

- Ising model, small 20×20 lattice
- Simulate first a realisation from the Ising model
- Compute the likelihood function $l(\beta)$ for this realisation
- Results for $\varepsilon = 0, 10^{-5}, 10^{-4}, 10^{-3}, 10^{-2}$ and 10^{-1}

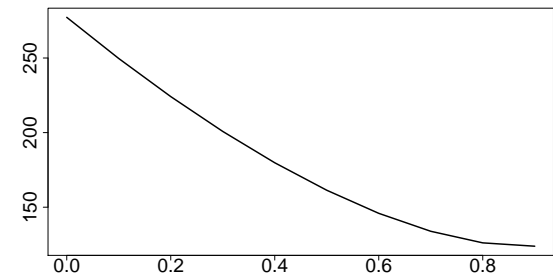
$\beta = 0.6$



$\beta = 0.8$



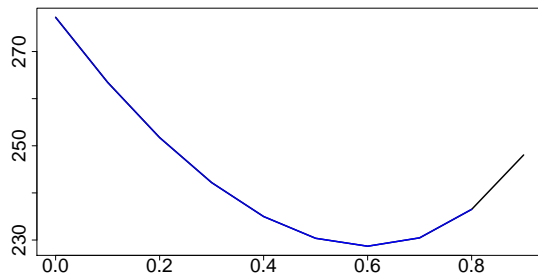
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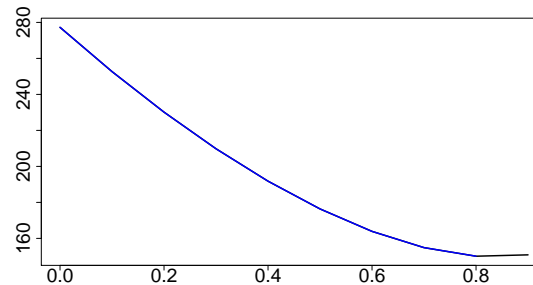
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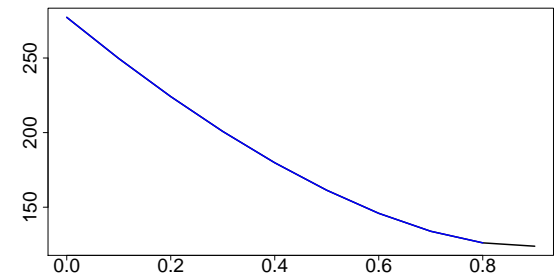
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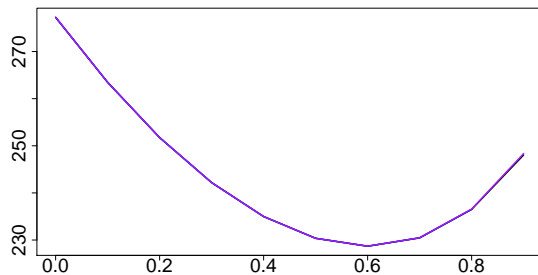
$\beta = 0.9$



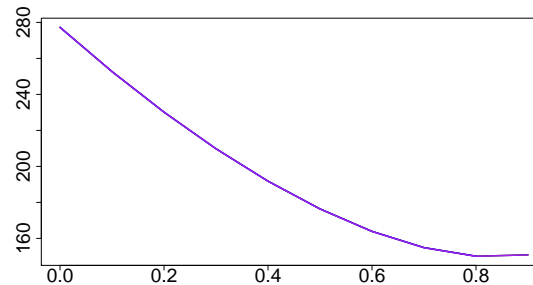
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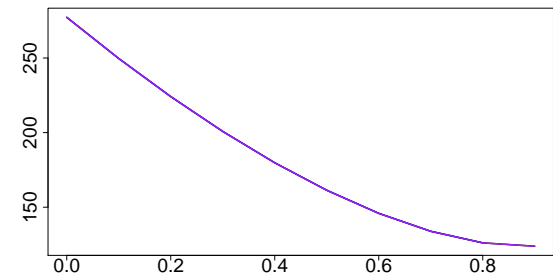
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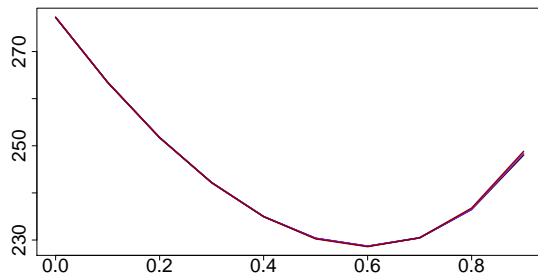
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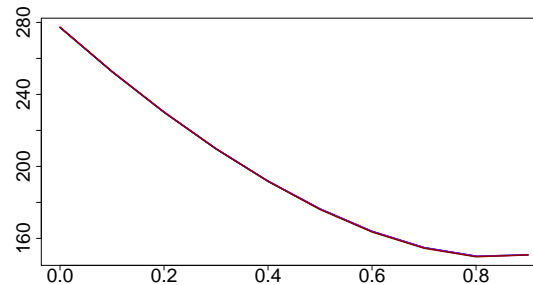
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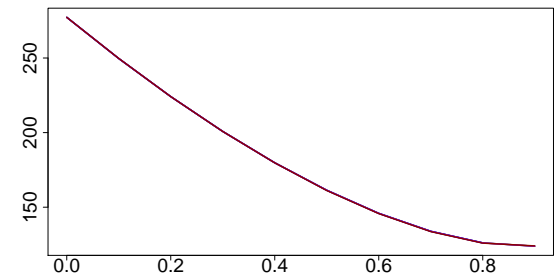
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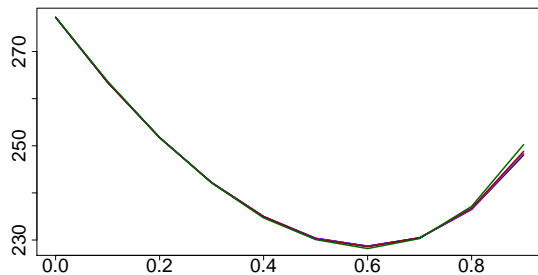
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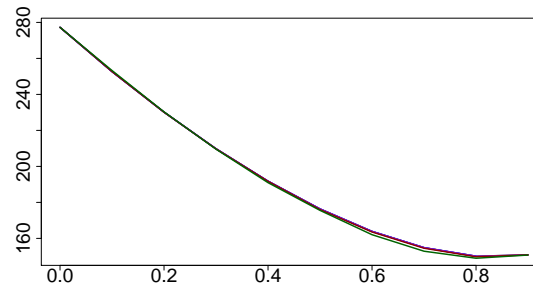
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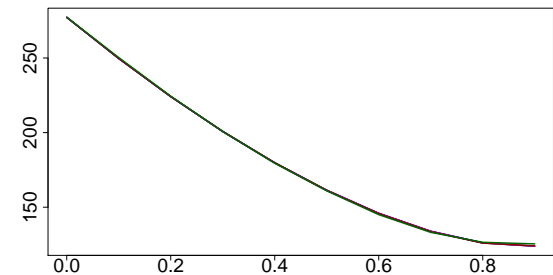
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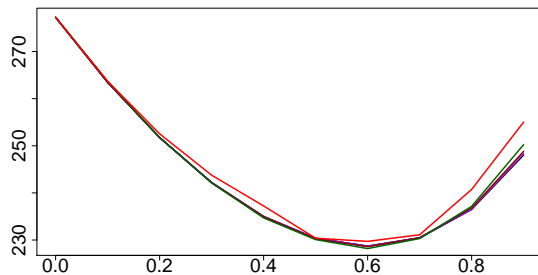
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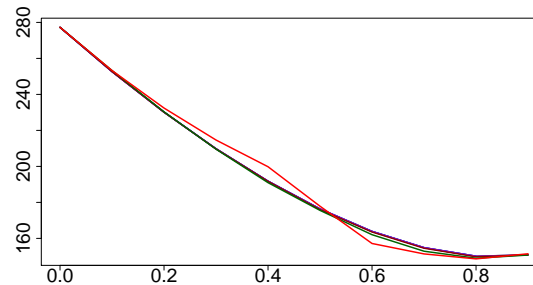
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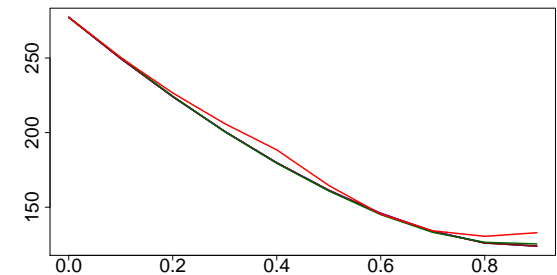
$\beta = 0.6$



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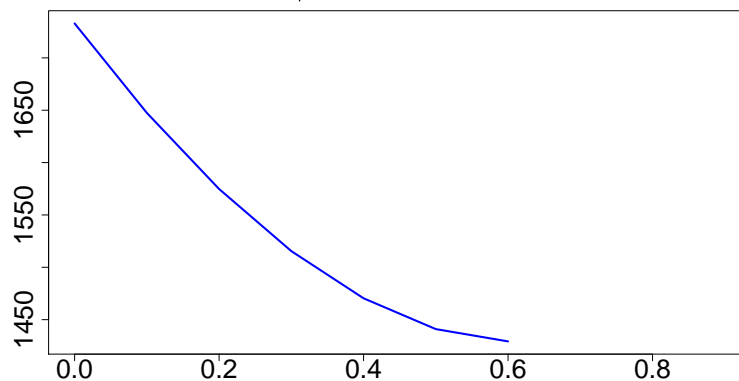
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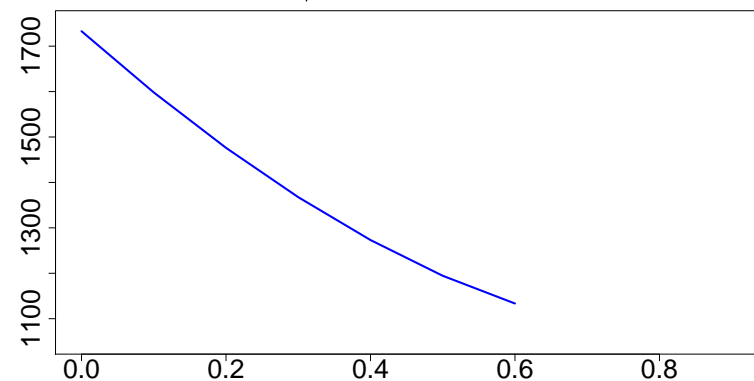
Compare results for different cut off value ε

- Ising model, 50×50 lattice
- Simulate first a realisation from the Ising model
- Compute the likelihood function $l(\beta)$ for this realisation
- Results for $\varepsilon = 10^{-5}, 10^{-4}, 10^{-3}, 10^{-2}$ and 10^{-1}

$\beta = 0.6$



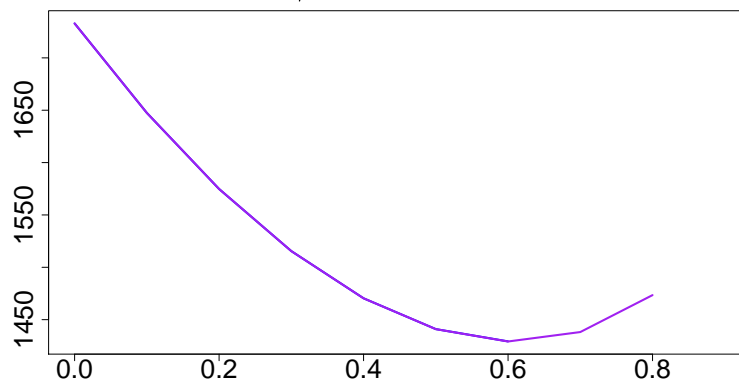
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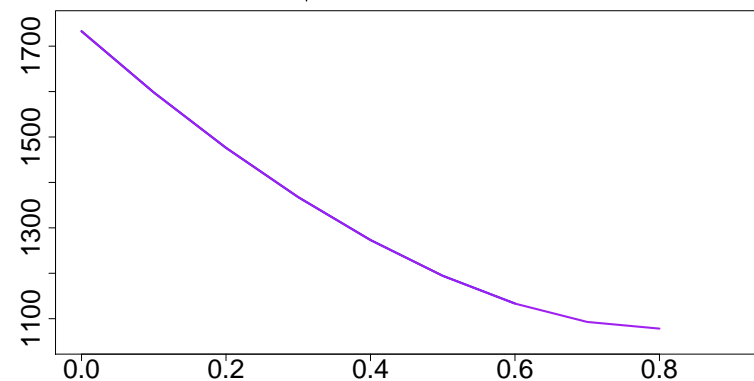
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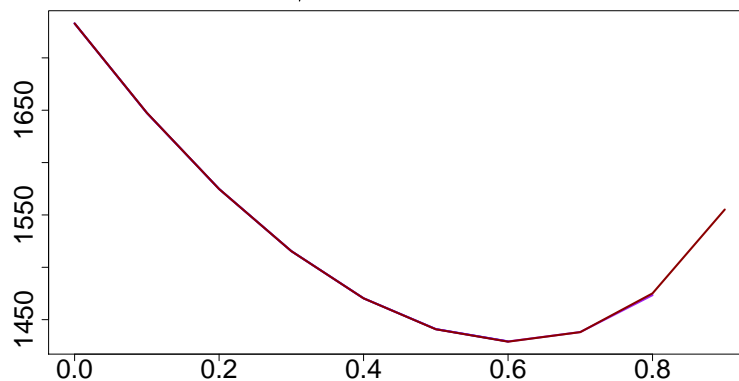
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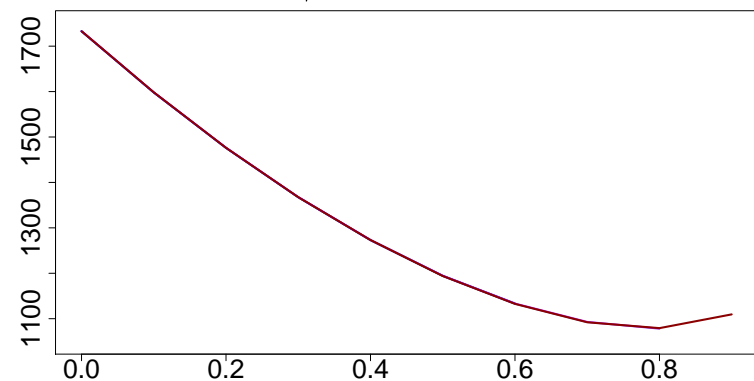
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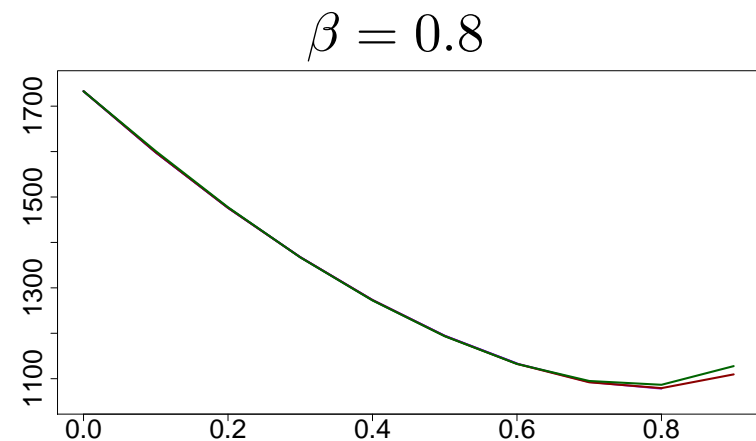
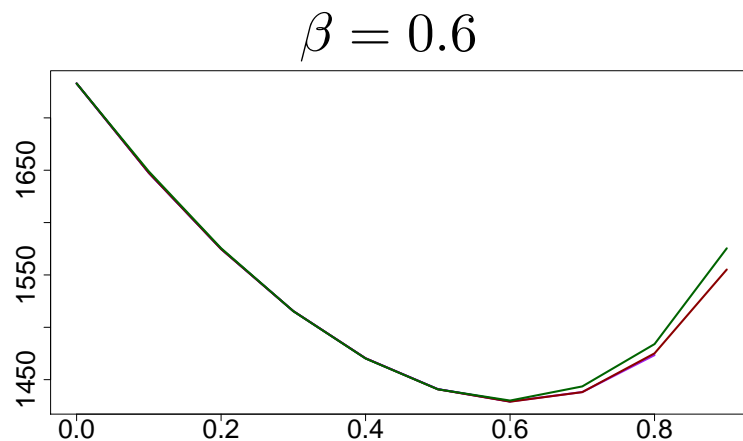


$\beta = 0.8$



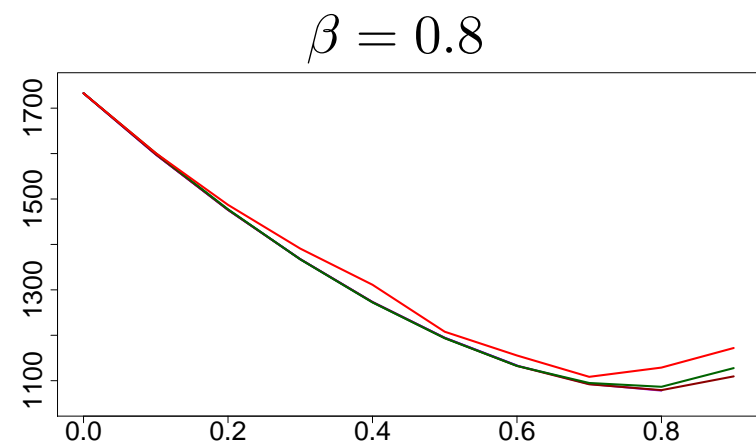
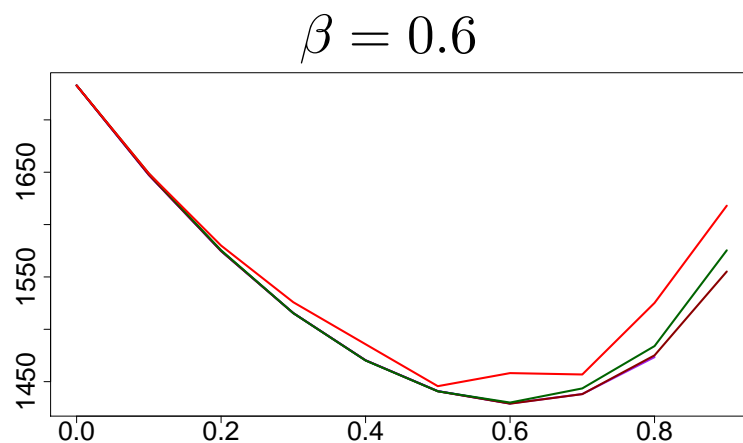
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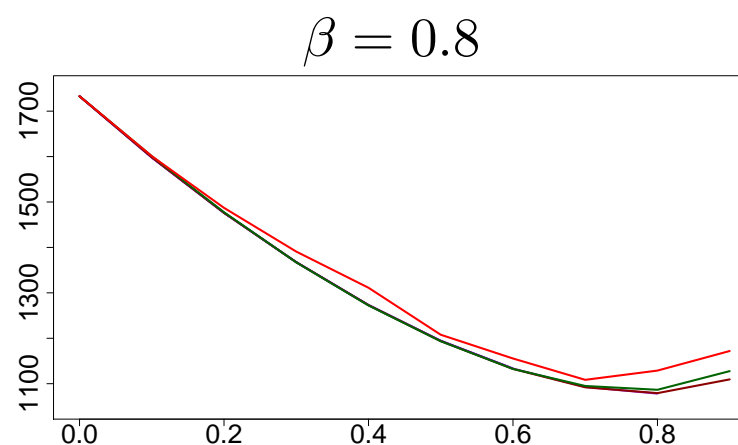
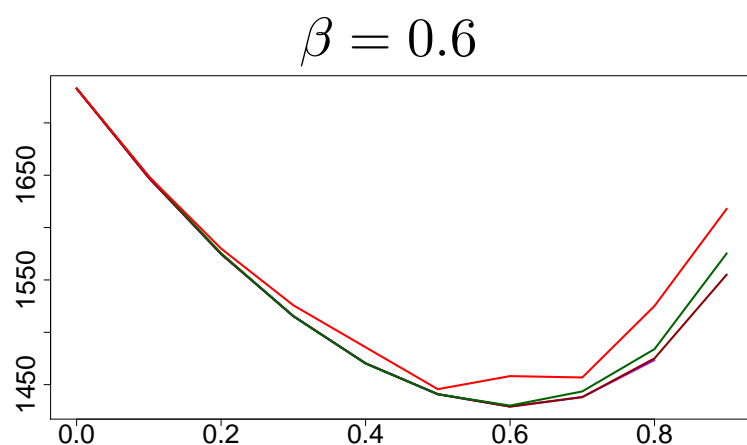
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- Fraction of interactions computed for $\varepsilon = 10^{-3}$

β	0.1	0.3	0.5	0.7	0.9
fraction	$2.1 \cdot 10^{-13}$	$5.0 \cdot 10^{-12}$	$7.0 \cdot 10^{-12}$	$3.3 \cdot 10^{-11}$	$1.5 \cdot 10^{-10}$

Used as proposal in a Metropolis–Hastings algorithm

- Ising model, 50×50 lattice
- Initiate Metropolis–Hastings algorithm with exact sample
- Run for $\varepsilon = 10^{-1}, 10^{-2}, 10^{-3}, 10^{-4}$ and 10^{-5}
- Monitor the Metropolis–Hastings acceptance rate

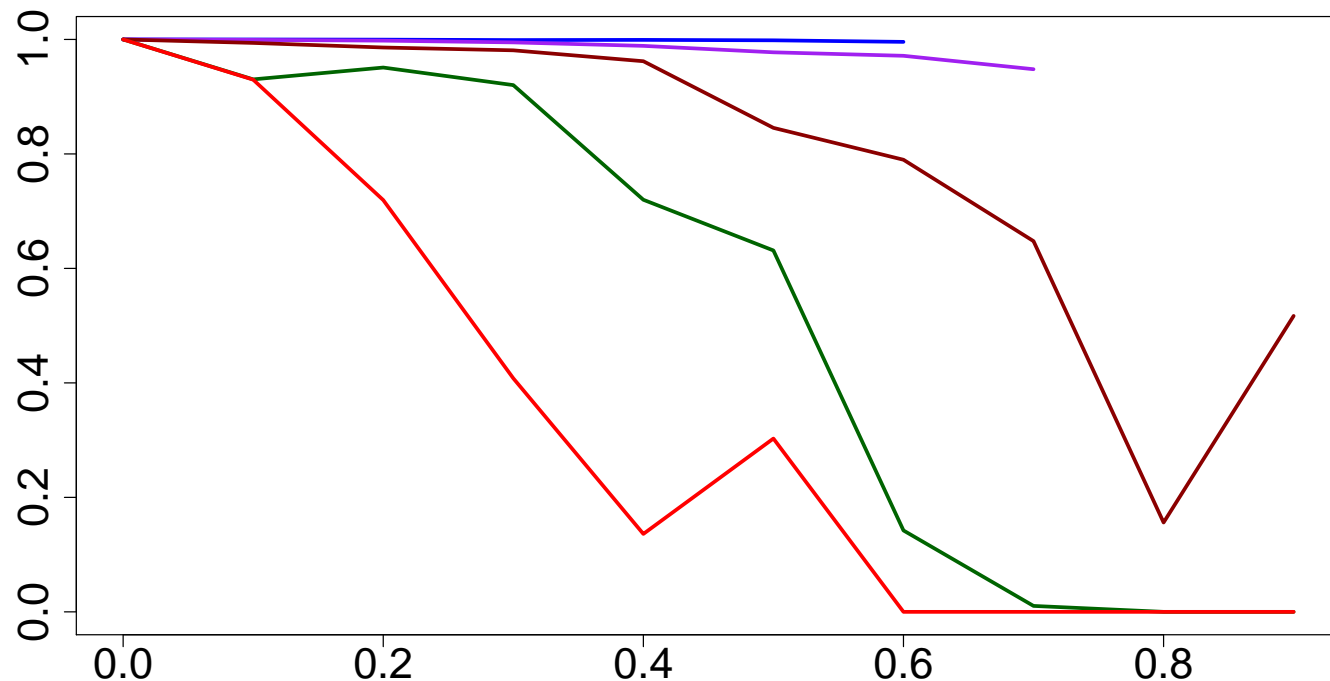


Image analysis example

- Ising prior, 50×50 lattice
- Conditionally independent Gaussian observations, $y_i|x_i \sim \mathbf{N}(x_i, \sigma^2)$
- Simulated data (true parameter values $\beta = 0.8, \sigma = 0.5$)
- Compute marginal likelihood function

$$l(\theta) = \pi(y|\theta) = \frac{\pi(x|\theta)\pi(y|x, \theta)}{\pi(x|y, \theta)}$$

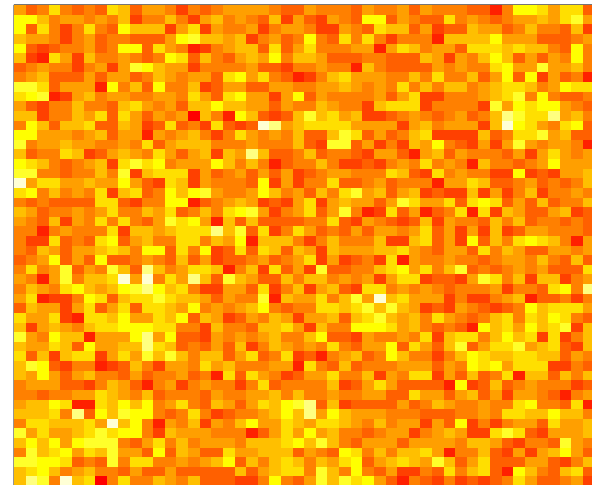
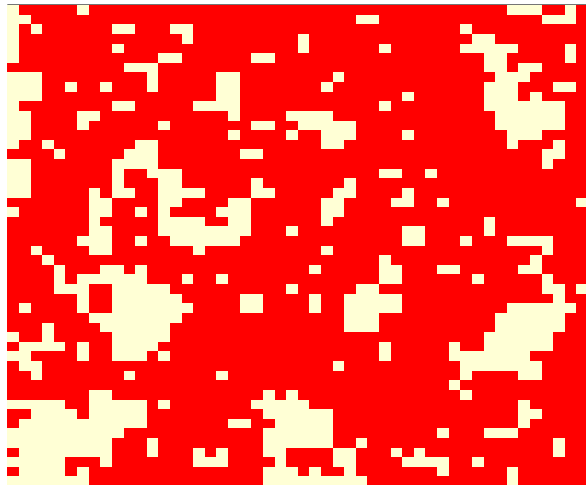


Image analysis example (cont.)

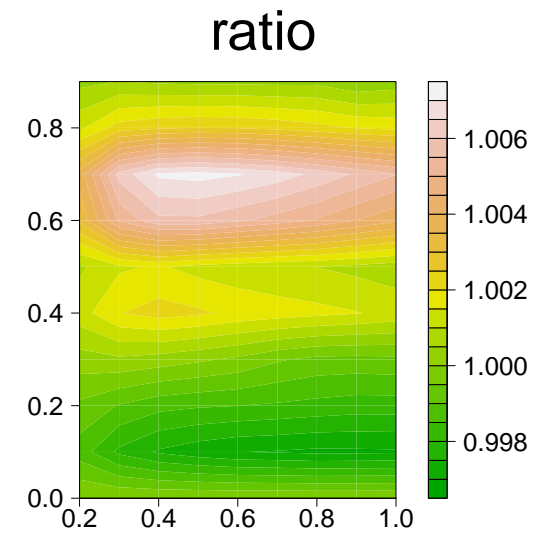
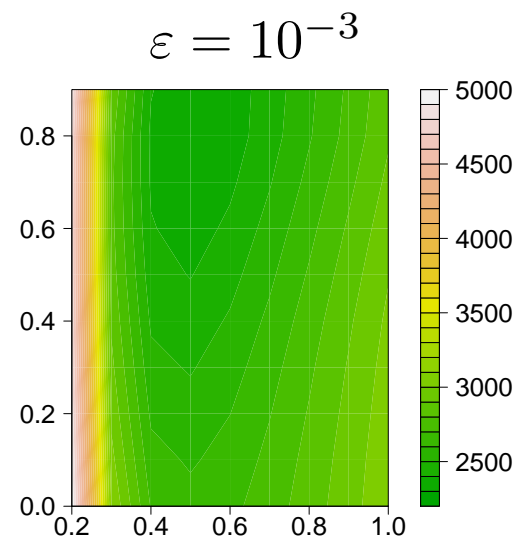
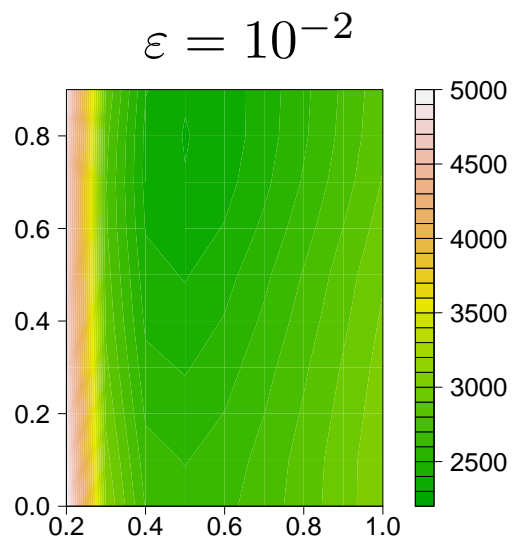
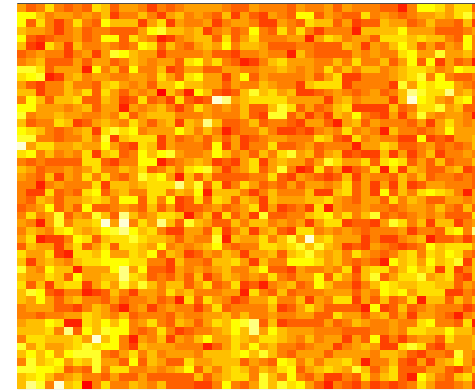
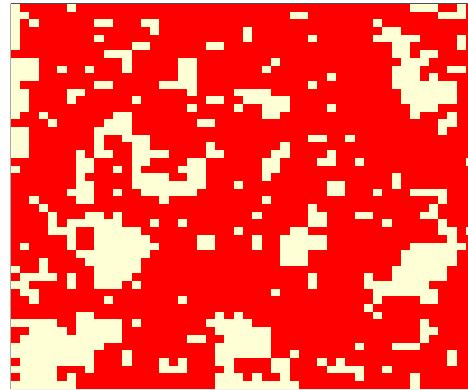
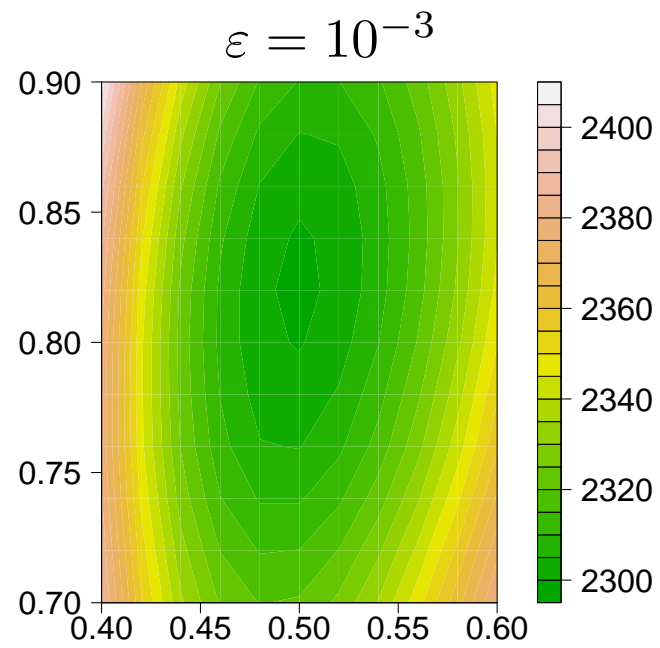
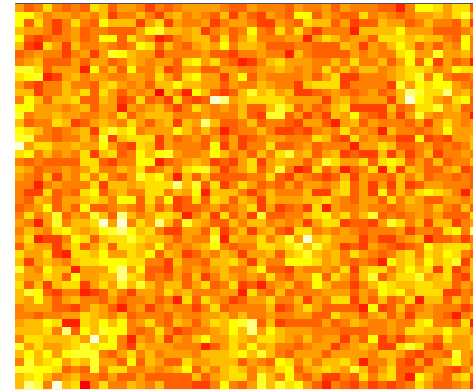
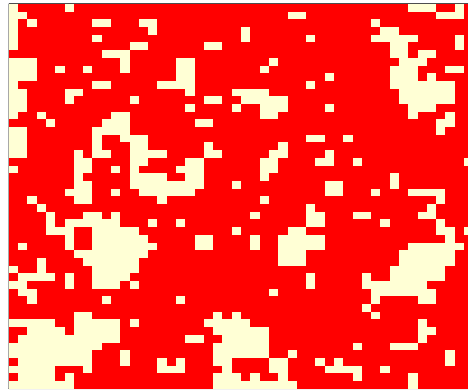
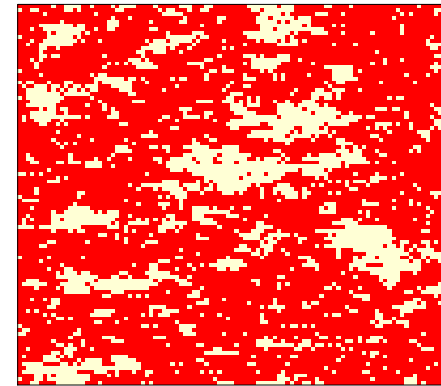
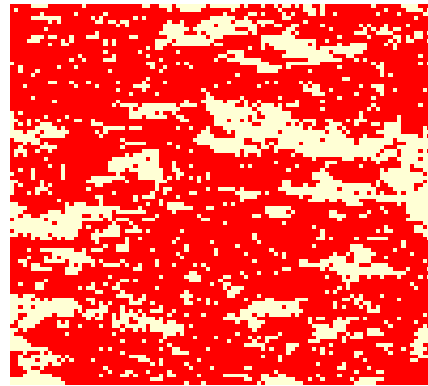
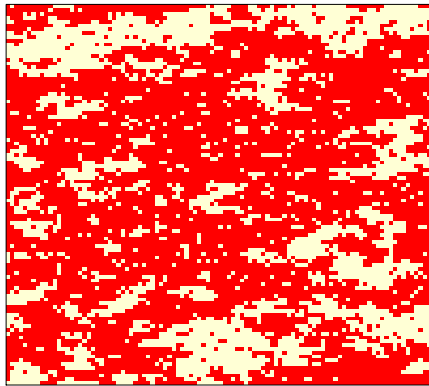


Image analysis example (cont.)

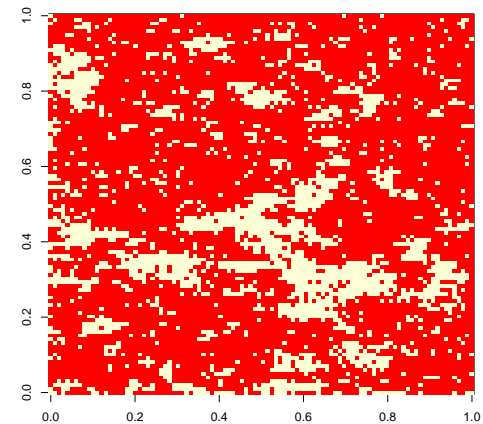
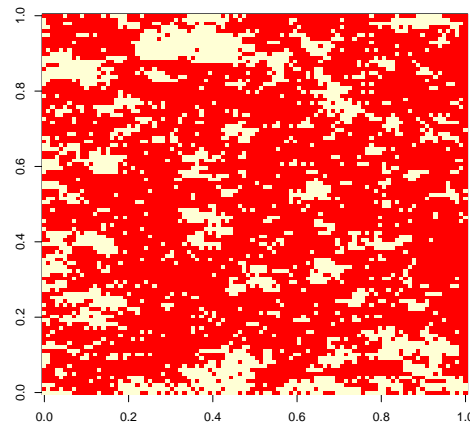
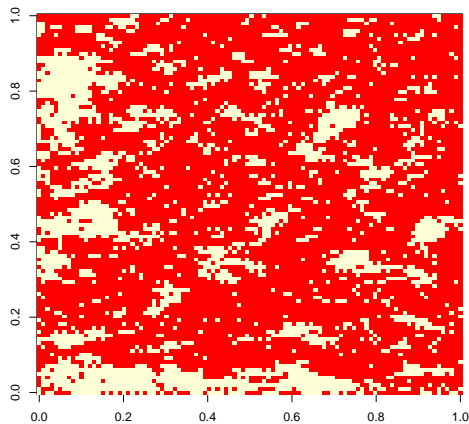


Example with 3×3 clique

- 3×3 clique model, pairwise interaction only, 100×100 lattice
- Realisations from specified model, by MCMC

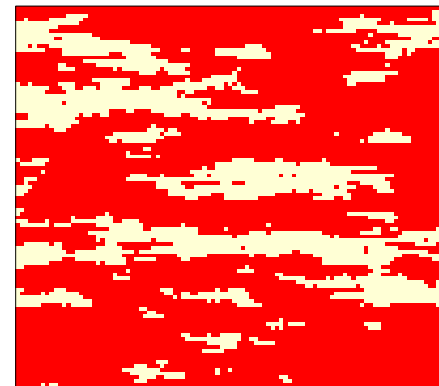
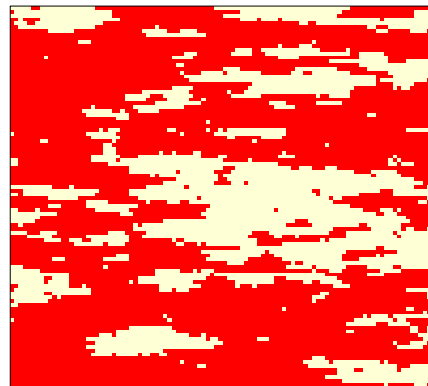
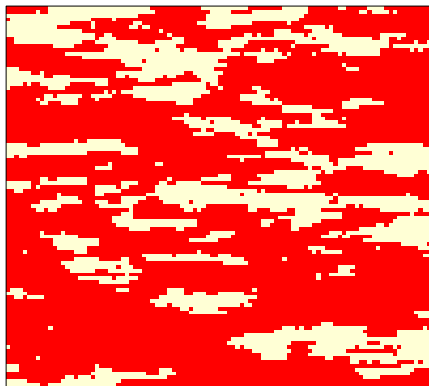


- Realisations from approximate model, $\varepsilon = 10^{-3}$



Problematic example

- 4×4 clique, strong higher order interactions
- Realisations from the model, by MCMC



- Too few interactions are approximated to zero — not able to run the approximation

A closer look at the approximate model

- Approximate model is

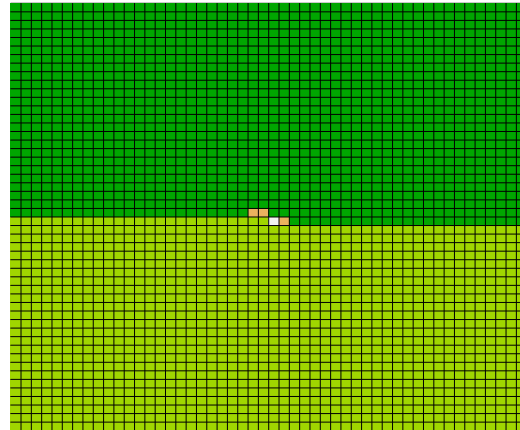
$$\tilde{\pi}(x_{1:n}|\theta) = \tilde{\pi}(x_n|\theta)\tilde{\pi}(x_{n-1}|x_n, \theta)\tilde{\pi}(x_{n-2}|x_{n-1:n}, \theta) \cdot \dots \cdot \tilde{\pi}(x_1|x_{2:n}, \theta)$$

- Conditionally independence because:
 - original MRF
 - approximation

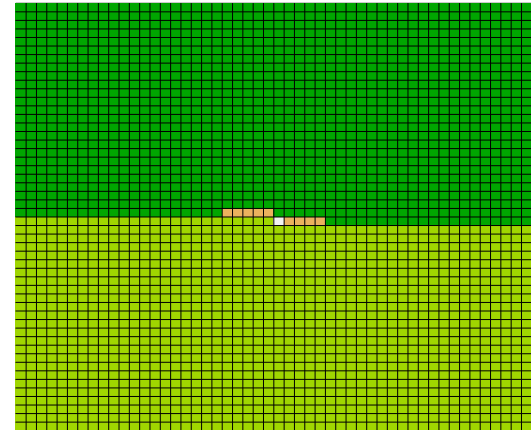
A closer look at the approximate model (cont.)

$\beta = 0.4$

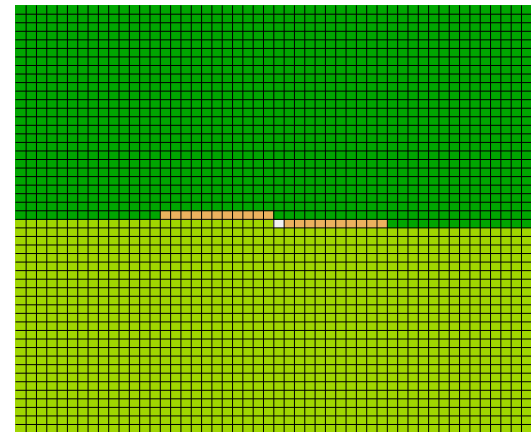
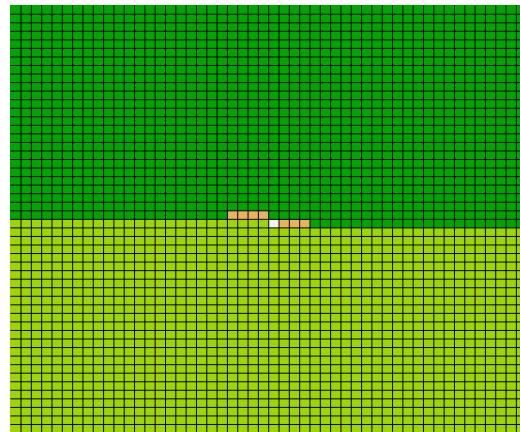
$\varepsilon = 10^{-2}$



$\varepsilon = 10^{-4}$



$\beta = 0.8$



Closing remarks

- For MRFs we have defined approximation to
 - the normalising constant — that (often) can be computed
 - the joint distribution — that (often) can be simulated directly
- The approximation is based on
 - representation of $b_k(x_{k:k+p}, \theta)$ by interaction parameters
 - most interaction parameters must be small
 - have assumed higher order interactions to be small whenever lower order interactions are small
- We have demonstrated the quality of the approximation in a number of examples
- The approximative model is a Markov mesh model (and POMM)
- Can sum out the variables in a different order