Image analysis

Generalized quad-trees

Random clusters

Percolation

Comparison

Simulation

Future work

References



Ising models and multiresolution quad-trees

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Image analysis

Generalized quad-trees

Random clusters

Percolation

Comparison

Simulation

Future work

References



Introduction

Multiresolution quad-trees are used in image analysis. For example: given a layer of pixels at finest resolution, successively aggregate blocks of pixels to produce coarser layers. A simple model then stipulates the black/white value of a pixel depends on its neighbours in the same layer and (with a different interaction strength) on its parent and daughters.



This talk describes initial steps in understanding the qualitative behaviour of this algorithm, addressing the following question:

"What do phase transitions look like when the finest layer of pixels is unconstrained ("free boundary")?"

For details, see Kendall and Wilson (2002).

Image analysis

Generalized quad-trees

Random clusters

Percolation

Comparison

Simulation

Future work

References



1. Image analysis

Wilson and Li (2002): Segmentation of noisy shapes. Bhalerao et al. (2001), Thönnes et al. (2002): application to MCMC in medical imaging.

Multiresolution MAP algorithm, 1.3% misclassification:







Quad-tree formed by successive averaging using "decimation":





Multiresolution: pros and cons

- + **FAST** since low resolution "steers" high resolution;
- + adapted to some kinds of **HIGH-LEVEL** objects;
- can produce "BLOCKY" reconstructions: resolution hierarchy mediates all spatial interactions.

Possible solution

Add further explicitly spatial interactions?

Image analysis

Generalized quad-trees

Random clusters

Percolation

Comparison

Simulation

Future work

References



2. Generalized quad-trees

Define \mathbb{Q}_d as graph whose vertices are cells of all dyadic tessellations of \mathbb{R}^d , with edges connecting each cell to its 2d neighbours, and also its parent (covering cell in tessellation at next resolution down) and its 2^d daughters (cells which it covers in next resolution up).



Case d = 1:

Neighbours at same level also are connected.

Remark: No spatial symmetry!



Further define:

• $\mathbb{Q}_{d;r}$ as subgraph of \mathbb{Q}_d at resolution levels of *r* or higher;



• $\mathbb{Q}_d(\mathbf{o})$ as subgraph formed by \mathbf{o} and all its descendants.



- **Remark:** there are many graph-isomorphisms between $\mathbb{Q}_{d;r}$ and $\mathbb{Q}_{d;s}$, with natural \mathbb{Z}^d -action;
- Remark: there are graph homomorphisms injecting Q(o) into itself, sending o to x ∈ Q(o) (*semi-transitivity*).

Image analysis

Generalized quad-trees

Random clusters

Percolation

Comparison

Simulation

Future work

References



Simplistic analysis

Define J_{λ} to be strength of neighbour interaction, J_{τ} to be strength of parent interaction. If $S_x = \pm 1$ then probability of configuration is proportional to $\exp(-H)$ where

$$H = -\frac{1}{2} \sum_{\langle x, y \rangle \in \mathcal{E}(G)} J_{\langle x, y \rangle} (S_x S_y - 1) , \qquad (1)$$

for $J_{\langle x,y\rangle} = J_{\lambda}, J_{\tau}$ as appropriate.

If $J_{\lambda} = 0$ then the free Ising model on $\mathbb{Q}_d(\mathbf{o})$ is a branching process (Preston 1977; Spitzer 1975); if $J_{\tau} = 0$ then the Ising model on $\mathbb{Q}_d(\mathbf{o})$ decomposes into sequence of *d*-dimensional classical (finite) Ising models. So we know there is a phase change at $(J_{\lambda}, J_{\tau}) = (0, \ln(5/3))$ (branching processes), and expect one when $\lambda = 0+$, indeed at $(J_{\lambda}, J_{\tau}) = (\ln(1 + \sqrt{2}), 0+)$ (2-dimensional Ising).

But is this all that there is to say?

Image analysis

Generalized quad-trees

Random clusters

Percolation

Comparison

Simulation

Future work

References



3. Random clusters

A similar problem, concerning Ising models on products of trees with Euclidean lattices, is treated by Newman and Wu (1990). We follow them by exploiting the celebrated *Fortuin-Kasteleyn random cluster representation* (Fortuin and Kasteleyn 1972; Fortuin 1972a; Fortuin 1972b):

The Ising model is the marginal site process at q = 2 of a site/bond process derived from a dependent bond percolation model with configuration probability $\mathbb{P}_{q,p}$ proportional to

 $q^C \times \prod_{\langle x,y \rangle \in \mathcal{E}(G)} \left((p_{\langle x,y \rangle})^{b_{\langle x,y \rangle}} \times (1 - p_{\langle x,y \rangle})^{1 - b_{\langle x,y \rangle}} \right) \,.$

(where $b_{\langle x,y \rangle}$ indicates whether or not $\langle x,y \rangle$ is closed, and C is the number of connected clusters of vertices). Site spins are chosen to be the same in each cluster independently of other clusters with equal probabilities for ± 1 .

WARWICK Image analysis Generalized quad-trees Random clusters Percolation Comparison Simulation Future work References Home Page Title Page 44 •• Page 10 of 2 Go Back Full Screen Close Quit

Random cluster facts

• (Representation of Ising model.) The marginal bond process is Ising with

$$p_{\langle x,y\rangle} = 1 - \exp(-J_{\langle x,y\rangle}); \qquad (2)$$

• (FK-comparison inequalities.) If $q \ge 1$ and A is an *increasing* event then

$$\mathbb{P}_{q,p}(A) \leq \mathbb{P}_{1,p}(A) \tag{3}$$

$$\mathbb{P}_{q,p}(A) \geq \mathbb{P}_{1,p'}(A) \tag{4}$$

where

$$p'_{\langle x,y
angle} = rac{p_{\langle x,y
angle}}{p_{\langle x,y
angle} + (1 - p_{\langle x,y
angle})q} = rac{p_{\langle x,y
angle}}{q - (q - 1)p_{\langle x,y
angle}} \,.$$

Since $\mathbb{P}_{1,p}$ is bond percolation (bonds open or not independently of each other), we can find out about phase transitions by studying *independent* bond percolation.

Image analysis

Generalized quad-trees

Random clusters

Percolation

Comparison

Simulation

Future work

References



4. Percolation

Independent bond percolation on products of trees with Euclidean lattices have been studied by Grimmett and Newman (1990), and these results were used in the Newman and Wu work on the Ising model. So we can make good progress by studying independent bond percolation on \mathbb{Q}_d , using p_{τ} for parental bonds, p_{λ} for neighbour bonds.

Theorem 1 There is almost surely no infinite cluster in $\mathbb{Q}_{d;0}$ (and consequently in $\mathbb{Q}_d(\mathbf{o})$) if

$$2^d au \mathcal{X}_{\lambda} \left(1 + \sqrt{1 - \mathcal{X}_{\lambda}^{-1}} \right) < 1,$$

where \mathcal{X}_{λ} is the mean size of the percolation cluster at the origin for λ -percolation in \mathbb{Z}^d .

Modelled on Grimmett and Newman (1990, §3 and §5). Get $\left(1 + \sqrt{1 - \mathcal{X}_{\lambda}^{-1}}\right)$ from matrix spectral asymptotics.



Image analysis

Generalized quad-trees

Random clusters

Percolation

Comparison

Simulation

Future work

References



Case of small τ

Need d = 2 for mathematical convenience. Use Borel-Cantelli argument and planar duality to show, for supercritical $\lambda > 1/2$ (that is, supercritical with respect to planar bond percolation!), all but finitely many of the resolution layers $L_n = [1, 2^n] \times [1, 2^n]$ of $\mathbb{Q}_2(\mathbf{o})$ have just one large cluster each of diameter larger than constant $\times n$.

Hence ...

Theorem 2 When $\lambda > 1/2$ and τ is positive there is one and only one infinite cluster in $\mathbb{Q}_2(\mathbf{o})$.





The story so far: adds small τ for case d = 2.

Image analysis

Generalized quad-trees

Random clusters

Percolation

Comparison

Simulation

Future work

References



Uniqueness of infinite clusters

The Grimmett and Newman (1990) work was remarkable in pointing out that as τ increases so there is a *further* phase change, from many to just one infinite cluster for $\lambda > 0$. The work of Grimmett and Newman carries through for $\mathbb{Q}_d(\mathbf{o})$. However the relevant bound is *improved* by a factor of $\sqrt{2}$ if we take into account the hyperbolic structure of $\mathbb{Q}_d(\mathbf{o})$!

Theorem 3 If $\tau < 2^{-(d-1)/2}$ and $\lambda > 0$ then there cannot be just one infinite cluster in $\mathbb{Q}_{d;0}$.

Method: sum weights of "up-paths" in $\mathbb{Q}_{d;0}$ starting, ending at level 0. For fixed *s* and start point there are infinitely many such uppaths containing *s* λ -bonds; but no more than $(1 + 2d + 2^d)^s$ which cannot be reduced by "shrinking" excursions. Hence control the mean number of open up-paths stretching more than a given distance at level 0.



Image analysis

Generalized quad-trees

Random clusters

Percolation

Comparison

Simulation

Future work

References



Contribution to upper bound on second phase transition:

Theorem 4 If $\tau > \sqrt{2/3}$ then the infinite cluster of $\mathbb{Q}_{2:0}$ is almost surely unique for all positive λ .

Method: prune bonds, branching processes, 2-dim comparison ...







The story so far: includes uniqueness transition for case d = 2.

Image analysis

Generalized quad-trees

Random clusters

Percolation

Comparison

Simulation

Future work

References



5. Comparison

We need to apply the Fortuin-Kasteleyn comparison inequalities (3) and (4). The event "just one unique infinite cluster" is **not** increasing, so we need more. Newman and Wu (1990) show it suffices to establish a *finite island property* for the site percolation derived under adjacency when all infinite clusters are removed. Thus:





Image analysis Generalized quad-trees Random clusters

Percolation

Comparison

Simulation

Future work

References



6. Simulation

Approximate simulations confirm the general story:

http://www.dcs.warwick.ac.uk/~rgw/sira/sim.html

- (1) Only 200 resolution levels;
- (2) At each level, 1000 sweeps in scan order;
- (3) At each level, simulate square sub-region of 128 × 128 pixels conditioned by mother 64 × 64 pixel region;
- (4) Impose periodic boundary conditions on 128×128 square region;
- (5) At the coarsest resolution, all pixels set white. At subsequent resolutions, 'all black' initial state.

Image analysis

Generalized quad-trees

Random clusters

Percolation

Comparison

Simulation

Future work

References





(a) $J_{\lambda} = 1, J_{\tau} = 0.5$

(d) $J_{\lambda} = 0.5, J_{\tau} = 0.5$

(b)
$$J_{\lambda}=1, J_{ au}=1$$

(c) $J_{\lambda} = 1, J_{\tau} = 2$



(g) $J_{\lambda} = 0.25, J_{\tau} = 0.5$ (h) $J_{\lambda} = 0.25, J_{\tau} = 1$ (i) $J_{\lambda} = 0.25, J_{\tau} = 2$

WARWICK Image analysis Generalized quad-trees Random clusters Percolation Comparison Simulation Future work References Home Page Title Page 44 •• Page 22 of 2 Go Back Full Screen Close Quit

7. Future work

This is about the *free* Ising model on $\mathbb{Q}_2(\mathbf{o})$. Image analysis more naturally concerns the case of prescribed boundary conditions (say, image at finest resolution level ...).

Question: will boundary conditions at "infinite fineness" propagate back to finite resolution?

Series and Sinaĭ (1990) show answer is yes for analogous problem on hyperbolic disk (2-dim, all bond probabilities the same). Gielis and Grimmett (2001) point out (eg, in \mathbb{Z}^3 case) these boundary conditions translate to a *conditioning* for random cluster model, and investigate using large deviations.

Project: do same for $\mathbb{Q}_2(\mathbf{o}) \dots$ and get quantitative bounds?

Image analysis Generalized quad-trees Random clusters Percolation Comparison Simulation Future work References



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Image analysis

- Generalized quad-trees
- Random clusters
- Percolation
- Comparison
- Simulation
- Future work
- References



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Image analysis Generalized quad-trees Random clusters

Percolation

Comparison

Simulation

Future work

References



A. Notes on proof of Theorem 1

Mean size of cluster at o bounded above by

$$\begin{split} &\sum_{n=0}^{\infty} \sum_{\underline{\mathbf{t}}:|\underline{\mathbf{t}}|=n} \mathcal{X}_{\lambda} \tau^{n} (\mathcal{X}_{\lambda}-1)^{T(\underline{\mathbf{t}})} \mathcal{X}_{\lambda}^{n-T(\underline{\mathbf{t}})} \\ &\leq \sum_{n=0}^{\infty} \mathcal{X}_{\lambda} (\tau \mathcal{X}_{\lambda})^{n} \sum_{\underline{\mathbf{t}}:|\underline{\mathbf{t}}|=n} (1-\mathcal{X}_{\lambda}^{-1})^{T(\underline{\mathbf{t}})} \\ &\leq \sum_{n=0}^{\infty} \mathcal{X}_{\lambda} (2^{d} \tau \mathcal{X}_{\lambda})^{n} \sum_{\underline{\mathbf{j}}:|\underline{\mathbf{j}}|=n} (1-\mathcal{X}_{\lambda}^{-1})^{T(\underline{\mathbf{j}})} \\ &\approx \sum_{n=0}^{\infty} \mathcal{X}_{\lambda} (2^{d} \tau \mathcal{X}_{\lambda})^{n} \left(1+\sqrt{1-\mathcal{X}_{\lambda}^{-1}}\right)^{n} \,. \end{split}$$

For last step, use spectral analysis of matrix representation

$$\sum_{\mathbf{\underline{j}}:|\mathbf{\underline{j}}|=n} (1 - \mathcal{X}_{\lambda}^{-1})^{T(\mathbf{\underline{j}})} = \begin{bmatrix} 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 - \mathcal{X}_{\lambda}^{-1} & 1 \end{bmatrix}^{n} \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Image analysis Generalized quad-trees Random clusters Percolation

Comparison

Simulation

Future work

References



B. Notes on proof of Theorem 2

Uniqueness: For negative exponent $\xi(1-\lambda)$ of dual connectivity function, set

 $\ell_n = (n \log 4 + (2 + \epsilon) \log n) \xi (1 - \lambda).$

More than one " ℓ_n -large" cluster in L_n forces existence of open path in dual lattice longer than ℓ_n . Now use Borel-Cantelli

On the other hand super-criticality will mean *some* distant points in L_n are inter-connected.

Existence: consider $4^{n-[n/2]}$ points in L_{n-1} and specified daughters in L_n . Study probability that

(a) parent percolates more than ℓ_{n-1} ,

- (b) parent and child are connected,
- (c) child percolates more than ℓ_n .

Now use Borel-Cantelli again

Image analysis Generalized quad-trees Random clusters Percolation

Comparison

Simulation

Future work

References



C. Notes on proof of Theorem 3

Two relevant lemmas:

Lemma 1 Consider $\boldsymbol{u} \in L_{s+1} \subset \mathbb{Q}_d$ and $\boldsymbol{v} = \mathcal{M}(\boldsymbol{u}) \in L_s \subset \mathbb{Q}_d$. There are exactly 2^d solutions in L_{s+1} of

$$\mathcal{M}(\mathbf{x}) = \mathcal{S}_{\mathbf{u};\mathbf{v}}(\mathbf{x}).$$

One is $\mathbf{x} = \mathbf{u}$. The others are the remaining $2^d - 1$ vertices \mathbf{y} such that the closure of the cell representing \mathbf{y} intersects the vertex shared by the closures of the cells representing \mathbf{u} and $\mathcal{M}(\mathbf{u})$. Finally, if $\mathbf{x} \in L_{s+1}$ does not solve $\mathcal{M}(\mathbf{x}) = S_{\mathbf{u};\mathbf{v}}(\mathbf{x})$ then

$$|\mathcal{S}_{\boldsymbol{u};\boldsymbol{v}}(\boldsymbol{x}) - \mathcal{S}_{\boldsymbol{u};\boldsymbol{v}}(\boldsymbol{u})||_{s,\infty} > \|\mathcal{M}(\boldsymbol{x}) - \mathcal{M}(\boldsymbol{u})\|_{s,\infty}.$$
(5)

Lemma 2 Given distinct v and y in the same resolution level. Count pairs of vertices u, x in the resolution level one step higher, such that

(a)
$$\mathcal{M}(\boldsymbol{u}) = \boldsymbol{v}$$
; (b) $\mathcal{M}(\boldsymbol{x}) = \boldsymbol{y}$; (c) $\mathcal{S}_{\boldsymbol{u};\boldsymbol{v}}(\boldsymbol{x}) = \boldsymbol{y}$.

There are at most 2^{d-1} such vertices.

Image analysis Generalized quad-trees Random clusters

Percolation

Comparison

Simulation

Future work

References



D. Notes on proof of Theorem 4

Prune! Then a direct connection is certainly established across the boundary between the cells corresponding to two neighbouring vertices \mathbf{u} , \mathbf{v} in L_0 if

(a) the τ -bond leading from **u** to the relevant boundary is open;

- (b) a τ -branching process (formed by using τ -bonds mirrored across the boundary) survives indefinitely, where this branching process has family-size distribution Binomial $(2, \tau^2)$;
- (c) the τ -bond leading from **v** to the relevant boundary is open.

Then there are infinitely many chances of making a connection across the cell boundary.



Image analysis Generalized quad-trees Random clusters Percolation

Comparison

Simulation

Future work

References



E. Notes on proof of infinite island property

Notion of "cone boundary" $\partial_c(S)$ of finite subset S of vertices: collection of daughters **v** of S such that $\mathbb{Q}_d(\mathbf{v}) \cap S = \emptyset$.

Use induction on S, building it layer L_n on layer L_{n-1} to obtain an isoperimetric bound: $\#(\partial_c(S)) \ge (2^d - 1) \#(S)$. Hence deduce

 $\mathbb{P}[S \text{ in island at } \mathbf{u}] \leq (1 - p_{\tau}(1 - \eta))^{(2^d - 1)n}$

where #(S) = n and $\eta = \mathbb{P}[\mathbf{u} \text{ not in infinite cluster of } \mathbb{Q}_d(\mathbf{u})].$

Upper bound on number N(n) of self-avoiding paths S of length n beginning at \mathbf{u}_0 :

$$N(n) \leq (1+2d+2^d)(2d+2^d)^n$$
.

Hence upper bound on the mean size of the island:

$$\sum_{n=0}^{\infty} (1+2d+2^d)(2d+2^d)^n \eta_{\rm br}^{n(1-2^{-d})},$$

where $\eta_{\rm br}$ is extinction probability for branching process based on Binomial $(2^d, p_{\tau})$ family distribution.