University of
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Percolation

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2.1 Initial Definitions

We start with some basic definitions for Percolation on cubic lattices, specifically bond percolation where we consider the edges on the graph to be either open or closed.

Definition 2.1. $\mathbb{Z} = \{..., -2, -1, 0, 1, 2, ...\}$ and $\mathbb{Z}^d = \{(x_1, x_2, ..., x_d) : x_i \in \mathbb{Z}\}$

Definition 2.2. For $x, y \in \mathbb{Z}^d$, define the distance from x to y, denoted $\delta(x, y)$, by

$$\delta(x, y) := \sum_{i=1}^{d} |x - y|$$

Definition 2.3 (d-dimensional cubic lattice)**.** We construct the lattice with vertices in $\mathbb{Z}^d$ and edges where the distance between vertices is one.

$$E(\mathbb{Z}^d) = \{\{u, v\} : u, v \in V(\mathbb{Z}^d), \ \delta(u, v) = 1\}$$

We will often refer to this lattice by the vertex set $\mathbb{Z}^d$ without specifying the edge set. We also denote the origin by 0.

2.2 Probability Space

We now introduce the Measure theory basics required to define the probability measure and subsequently the probability space for our percolation model.

Definition 2.4 (σ-algebra)**.** For some set X, we call $\mathcal{A} \subseteq \mathcal{P}(X)$, a subset of the power set of X, a σ-algebra of X if:

1. $\emptyset, X \in \mathcal{A}$

2. $A \in \mathcal{A} \implies A^c := X \setminus A \in \mathcal{A}$ (Closed under complement)

3. For $A_i \in \mathcal{A}, i \in \mathbb{N}$ we have that $\bigcup_{i=1}^{\infty} A_i \in \mathcal{A}$ (Closed under countable unions)

We call the pair $(X, \mathcal{A})$ a measurable space, and elements of $\mathcal{A}$ measurable sets.

We note by De Morgan's Laws that a σ-algebra is also closed under countable intersections.

Definition 2.5 (Measure)**.** A measure μ on a measurable space $(X, \mathcal{A})$ is a function $\mu : \mathcal{A} \to [0, \infty]$ where $\mu(\emptyset) = 0$ and for disjoint $A_i \in \mathcal{A}, i \in \mathbb{N}$ we have that

$$\mu(\bigcup_{i=1}^{\infty} A_i) = \sum_{i=1}^{\infty} \mu(A_i)$$

We call the triple $(X, \mathcal{A}, \mu)$ a measure space.

In the context above if we think about X being our set of outcomes, and the sigma algebra of X representing the set of events we wish to assign a probability to, it's intuitive to give these events a probability by a measure. Clearly we need to restrict our probability measure such that its domain is the interval $[0, 1]$

Definition 2.6 (Probability Measure). Let Ω be the set of all outcomes (the sample space) and $\mathcal{F}$ be a σ-algebra of Ω where the elements are events we wish to consider (the event space). Then a measure $\mathbb{P}$ on $(\Omega, \mathcal{F})$ is a probability measure if:

1. $\mathbb{P} : \mathcal{F} \rightarrow [0, 1]$

2. $\mathbb{P}(\Omega) = 1$

Then we call the triple $(\Omega, \mathcal{F}, \mathbb{P})$ a probability space.

2.3   The Model

We now take some $p \in [0, 1]$ which will be our parameter which specifies the probability a given edge is open. Setting $q = 1 - p$ we say each edge is independently open with probability p and closed with probability q. We can think of the open and closed edges defining a random subgraph of $\mathbb{Z}^d$ where only edges set to open are retained.

We let our sample space $\Omega = \prod_{e \in E}\{0, 1\}$ where $E = E(\mathbb{Z}^d)$ and an edge in state 1 represents it is open, and 0 that it is closed.
We may refer to Ω as the set of configurations, where a configuration ω is a function assigning edges to be open or closed, meaning

$$\omega : E \rightarrow \{0, 1\}, e \mapsto \omega_e$$

Then we must define $\mathcal{F}$, some σ-algebra of Ω, the events we wish to assign a probability to. Clearly we cannot just take $\mathcal{F} = \mathcal{P}(\Omega)$, since we have uncountably many configurations in Ω. This can be easily seen via a diagonalization argument. It turns out we can generate the σ-algebra we want by the cylinder sets of Ω. We base our definition on the one given by [2].

Definition 2.7 (Cylinder Set). We say a subset $S \subseteq \Omega$ is a cylinder set if and only if there exists a finite subset $F \subseteq E$ and $\sigma \in \{0, 1\}^F$ such that

$$S(F, \sigma) = \{\omega \in \Omega : \omega_f = \sigma_f \text{ for } f \in F\}$$

In essence the cylinder set is the set of configurations ω which map all the edges in F to the same state that σ does.

Then we define $\mathcal{F}$ to be the set of all unions of cylinder sets of Ω, then $\mathcal{F}$ is said to be generated by the cylinder sets of Ω. In this case $\mathcal{F}$ is a σ-algebra of Ω. Intuitively $\mathcal{F}$ is the set of events which only depend on a finite number of edges.

Next we define our probability measure on $(\Omega, \mathcal{F})$ by

$$\mathbb{P}_p(S(F, \sigma)) = \prod_{f \in F} (p(\sigma_f) + q(1 - \sigma_f))$$

Where as usual $q = 1 - p$, and we use the subscript p to emphasize that p is the parameter in our model. Notice that we have defined $\mathbb{P}_p$ for a cylinder sets rather than elements of the σ-algebra, however it turns out by Carathéodory's extension theorem the probability measure above has a unique extension to the whole σ-algebra. For the details of this theorem see Theorem 1.3.10 in [1]. Then $(\Omega, \mathcal{F}, \mathbb{P}_p)$ is our probability space in which we examine the percolation model.

We now lay out some definitions specific for percolation.

Definition 2.8. Let $C(x)$ denote the open cluster (component) containing x, which is the set of vertices in $\mathbb{Z}^d$ which are connected to x by a path of open edges. We abbreviate the open cluster containing the origin $C(0)$ by C

Definition 2.9 (Increasing Event)**.** An event $A \subseteq \Omega$ is increasing if when $\omega \in A$ and

$$\forall e \in E(\mathbb{Z}^d),\ \omega_e = 1 \implies \omega'_e = 1$$

Then $\omega' \in A$.

We notice that the event $\{|C| = \infty\}$ is clearly an increasing event, since adding open edges to a configuration with an infinite connected cluster containing the origin will not remove the infinite cluster.

Definition 2.10 (Percolation function)**.** We define the percolation function $\theta(p)$ as follows

$$\theta(p) = \mathbb{P}_p(|C| = \infty)$$

In words the percolation function is simply the probability that we can reach an infinite number of vertices from the origin by open edges. Furthermore, we also note that this is the same as asking what is the probability of having an infinite length self-avoiding path of open edges starting at the origin.

We intend to show that this percolation function is non-decreasing, but first we show a more general result for increasing events.

Lemma 2.11. *If $A \subseteq \Omega$ is an increasing event then $\mathbb{P}_p(A)$ is non-decreasing in p.*

Proof. We use the coupling of percolation processes to show that $\mathbb{P}_p(A)$ is non-decreasing. We use the definition of couplings given by [6].

Definition 2.12 (Coupling). Let $\mathbb{P}$ and $\mathbb{P}'$ be probability measures on the same measurable space $(\Omega, \mathcal{F})$. A coupling of $\mathbb{P}$ and $\mathbb{P}'$ is a probability measure $\mathbf{P}$ on $(\Omega \times \Omega, \mathcal{F} \times \mathcal{F})$ such that the marginals of $\mathbf{P}$ coincide with $\mathbb{P}$ and $\mathbb{P}'$. Meaning

$$\mathbf{P}(A \times \Omega) = \mathbb{P}(A) \quad \text{and} \quad \mathbf{P}(\Omega \times A) = \mathbb{P}'(A), \quad \forall A \in \mathcal{F}$$

We say (X, Y) is coupling of $\mathbb{P}$ and $\mathbb{P}'$ for random variables X, Y if the law of (X, Y) is a coupling of $\mathbb{P}$ and $\mathbb{P}'$ as defined above.

We construct a pair of configurations $\omega, \omega' \in \Omega$ using a family of uniform random variables $(U_e)_{e \in E}$ which are iid on [0,1]. Then fixing a pair $p, p' \in [0, 1]$ such that $p < p'$, for each $e \in E$ we define its state in ω by:

$$\omega_e = \begin{cases} 1 & \text{if } U_e \leq p \\ 0 & \text{otherwise} \end{cases}$$

So each edge in ω is independently open with probability p. Specifically the law or distribution of ω is $\mathbb{P}_p$ Similarly define ω' by:

$$\omega'_e = \begin{cases} 1 & \text{if } U_e \leq p' \\ 0 & \text{otherwise} \end{cases}$$

Then the law of ω'_e is $\mathbb{P}_{p'}$. Then since $p \leq p'$ it is clear that

$$\forall e \in E, \omega_e \leq \omega'_e$$

Which we denote by the shorthand $\omega \leq \omega'$.
Then (ω, ω') is a coupling of $\mathbb{P}_p$ and $\mathbb{P}_{p'}$ with the property that $\mathbf{P}(\omega \leq \omega') = 1$
Let some increasing event $A \subseteq \Omega$, by definition if $\omega \in A$, then $\omega' \in A$ so:

$$\mathbb{P}_p(A) = \mathbf{P}(\omega \in A) \leq \mathbf{P}(\omega' \in A) = \mathbb{P}_{p'}(A)$$

So we conclude $\mathbb{P}_p(A)$ is non-decreasing in p

$\square$

Corollary 2.13. *The percolation function $\theta(p)$ is non-decreasing in p.*

Proof. Since we already saw that the event $\{|C| = \infty\}$ is an increasing event, then by Lemma 2.11 we know that $\theta(p) = \mathbb{P}_p(|C| = \infty)$ is non-decreasing in p. $\square$

An interesting property of our percolation model is that it exhibits a phase transition. Where at some point the behaviour of the model changes notably. For the percolation model we see there is some value of our parameter p, before which we know almost surely there is no infinitely connected cluster, and after which there may be one. We call this value of p the critical value or percolation threshold.

Definition 2.14 (Critical Value)**.** We define the critical value $p_c(d)$ formally as follows

$$p_c(d) = \sup\{p \in [0,1] : \theta(p) = 0\}$$

where d denotes the dimensionality of our graph $\mathbb{Z}^d$, we may sometimes drop the d and just refer to p_c when it is clear what the model dimension is.

3 Existence of a critical value

3.1 Existence of a critical value on $\mathbb{Z}$

Trivially the critical value is p = 1. Consider the event $X_n = \{$There is an open self-avoiding path of length n starting at the origin$\}$ Then $X_n \supseteq X_{n+1}$ and so

$$\lim_{n\to\infty} \mathbb{P}_p(X_n) = \theta(p)$$

And since $\mathbb{P}_p(X_n) = 2p^n$, as the path can go left or right from the origin. We have for all $p < 1$, $\theta(p) = 0$. Thus, $\theta(p) > 0$ if and only if $p = 1$.

3.2 Existence of a critical value on $\mathbb{Z}^2$

We show the existence of the critical value in this case by bounding it from above and below. We follow the proofs given in [7]

Theorem 3.1. *If $p < 1/3$, $\theta(p) = 0$.*

Proof. Let $X_n = \{$There is an open self-avoiding path of length n starting at the origin$\}$ as in Section 3.1. Then the probability for a path of length n to be open on every edge is p^n. The number of paths of length n from the origin is at most $4(3^{n-1})$ since there are 4 edges to choose from at the origin, then for each next step in the path there are at most 3 edges we can pick as the path is self-avoiding. Hence, we get $\mathbb{P}_p(X_n) \leq 4(3^{n-1})p^n$. Then we take the limit since $\lim_{n\to\infty} \mathbb{P}_p(X_n) = \theta(p)$.

$$\lim_{n\to\infty} \mathbb{P}_p(X_n) \leq \lim_{n\to\infty} 4(3^{n-1})p^n$$
$$\leq 4 \cdot 3^{-1} \lim_{n\to\infty} (3p)^n$$

Since $p < 1/3$ we have $\theta(p) = \lim_{n\to\infty} \mathbb{P}_p(X_n) = 0$ $\qquad\qquad\square$

Theorem 3.2. *For p close to 1, we have $\theta(p) > 0$*

Proof. We introduce the dual graph $(\mathbb{Z}^2)^*$ which has vertices in $(\mathbb{Z}^2 + \binom{1/2}{1/2})$, and edges as you would expect between vertices at distance 1. Then we can see there is a clear correspondence between the edges of $\mathbb{Z}^2$ and its dual, since each edge in the dual intersects a unique edge in the original graph. Thus, we can create a mapping from the open and closed edges of $\mathbb{Z}^2$ to the dual graph, where the edge in the dual is open if and only if the intersecting edge in $\mathbb{Z}^2$ is open.

Then we notice that if there exists a cycle of closed edges in the dual graph enclosing the origin then the size of the open cluster at the origin is finite.

Lemma 3.3. $|C| < \infty \iff \exists$ *a cycle of closed edges in $(\mathbb{Z}^2)^*$ enclosing the origin*

Proof. Thinking visually since there is a ring of closed edges in the dual, the open edges from the origin in the original graph cannot extend beyond this ring. A more formal pure graph theory proof can be made but is omitted here. $\square$

Let $X_n = \{$There is a length n cycle of closed edges in $(\mathbb{Z}^2)^*$ which surrounds the origin$\}$ Then using Lemma 3.3 we see

$$\mathbb{P}_p(|C| < \infty) = \mathbb{P}_p(\bigcup_{n=4}^{\infty} \mathbb{P}_p(X_n)) \leq \sum_{n=4}^{\infty} \mathbb{P}_p(X_n) \leq \sum_{n=4}^{\infty} n \cdot 4(3^{n-1})q^n$$

Where in the last summation we get $4(3^{n-1})q^n$ as before by counting the number of possible edges a self avoiding walk can take then multiplying by the probability this path is open in the dual graph. We multiply this term by n since we know for every cycle around the origin in the dual there is some smallest $i \in \mathbb{N}$ for which the cycle crosses the edge $\{i, i-1\}$, i.e the positive x-axis. Then we can translate this cycle left and right to the another n positions, this gives up to n new cycles enclosing the origin in the dual graph.

This sum is finite when $q < 1/3$, which is when $p > 2/3$. We can make the sum arbitrarily small when $p \to 1$, when the sum is smaller than 1 this implies $\theta(p) > 0$ Taking the value $p = 0.9$ for example, we can calculate, using Wolfram Alpha in this case, that the sum converges to

$$\sum_{n=4}^{\infty} n \cdot 4(3^{n-1})(1 - 0.9)^n = 0.0683265$$

Thus $\theta(0.9) > 0$.

$\square$

Hence, the critical value $p_c(2) \in (\frac{1}{3}, 0.9)$

3.3 Existence of a critical value on $\mathbb{Z}^d$

We can easily derive the result that $p_c(d) \in (0,1)$ from the previous section. We notice that for all $d \geq 2$ we can extract $\mathbb{Z}^d$ from $\mathbb{Z}^{d+1}$ by simply ignoring the extra edges and their states. Thus, if there exists an infinite open cluster in $\mathbb{Z}^d$ this would also imply there is one in $\mathbb{Z}^{d+1}$. Since we have just shown $p_c(2) \in (0,1)$ then we must also have $p_c(d) \in (0,1)$ for all d.

TODO: explain critical value is non-decreasing in d, by coupling argument?

4 Harris-Kesten Theorem

An important result in percolation theory is that the critical value for the square lattice is $\frac{1}{2}$. This result came in two parts, firstly Harris [4] showed that $p_c(2) \leq \frac{1}{2}$ in 1960. Then 20 years later Kesten [5] proved $p_c(2) \geq \frac{1}{2}$, these proofs combined resulted in the following theorem.

Theorem 4.1 (Harris-Kesten theorem). *The critical value on $\mathbb{Z}^2$, $p_c(2) = \frac{1}{2}$.*

In this project our aim is to show the Harris part of the theorem, that $p_c(2) \leq \frac{1}{2}$. To do this we first introduce some more foundational results.

Recall from earlier the definition of an increasing event.

Definition 4.2 (Increasing Event). An event $A \subseteq \Omega$ is increasing if when $\omega \in A$ and

$$\forall e \in E(\mathbb{Z}^d), \omega_e = 1 \implies \omega'_e = 1$$

Then $\omega' \in A$.

We follow the definitions and proof for the Harris Inequality given by [3].

Theorem 4.3 (Harris Inequality). *If $A, B \subseteq \Omega$ are increasing events then*

$$\mathbb{P}(A \cap B) \geq \mathbb{P}(A)\mathbb{P}(B)$$

This is a specific case of a more general inequality, the FKG inequality. If f and g are two bounded increasing functions then

$$\mathbb{E}_p(fg) \geq \mathbb{E}_p(f)\mathbb{E}_p(g)$$

Proof. We show that the FKG inequality holds, since we can easily derive Harris' inequality by setting f and g to be indicator functions of the increasing events A and B. We first show the result holds for increasing functions which depend on a finite number of edges, then we see how this restriction

can be removed using the Martingale convergence theorem.

We work by induction on the number of edges these functions f and g depend on. Label the edges in $E = \{e_i : i \geq 1, e_i \in E\}$. We also shorten our notation for the state of a given edge $\omega_{e_i} = \omega_i$ for readability during this proof. In the base case $n = 1$ we suppose f and g depend solely on ω_1. We note that we can just show the inequality holds for $f(0) = g(0) = 0$ since additional constants in f and g will not affect the inequality. Since f and g are increasing we know that $f(1) \geq 0$ and $g(1) \geq 0$. So we find by the definition of expectation

$$\mathbb{E}_p(fg) - \mathbb{E}_p(f)\mathbb{E}_p(g) = (p \cdot f(1)g(1) + q \cdot 0) - (p \cdot f(1) + q \cdot 0)(p \cdot g(1) + q \cdot 0)$$

Then simplifying by removing terms multiplied by zero

$$\mathbb{E}_p(fg) - \mathbb{E}_p(f)\mathbb{E}_p(g) = pf(1)g(1) - p^2 f(1)g(1) \geq 0$$

Since $p, f(1), g(1) \geq 0$. So the result holds for $n = 1$.

Now assuming as our inductive hypothesis that the inequality holds for functions dependent on $n = k$ edges. We consider when $n = k + 1$. We fix the state of the first k edges, $\omega_1, ..., \omega_k$, and condition on them since we aim to use the tower law for expectation later. Then,

$$\mathbb{E}_p(fg|\omega_1, ..., \omega_k) = pf(\omega_1, ..., \omega_k, 1)g(\omega_1, ..., \omega_k, 1) + qf(\omega_1, ..., \omega_k, 0)g(\omega_1, ..., \omega_k, 0)$$

Where the functions $f(\omega_1, ..., \omega_k, 1)$ and $g(\omega_1, ..., \omega_k, 1)$ now depend only on the edge ω_{k+1} Then letting $\mathbb{P}_{\omega_{k+1}}$ be the law (distribution) of ω_{k+1} and similarly $\mathbb{E}_{\omega_{k+1}}$. We see that

$$\begin{aligned}
\mathbb{E}_p(fg|\omega_1, ..., \omega_k) &= pf(\omega_1, ..., \omega_k, 1)g(\omega_1, ..., \omega_k, 1) \\
&\quad + qf(\omega_1, ..., \omega_k, 0)g(\omega_1, ..., \omega_k, 0) \\
&= \mathbb{E}_{\omega_{k+1}}(f(\omega_1, ..., \omega_k, \cdot)g(\omega_1, ..., \omega_k, \cdot))
\end{aligned}$$

Then applying our inductive hypothesis.

$$\mathbb{E}_{\omega_{k+1}}(f(\omega_1, ..., \omega_k, \cdot)g(\omega_1, ..., \omega_k, \cdot)) \geq \mathbb{E}_{\omega_{k+1}}(f(\omega_1, ..., \omega_k, \cdot))\mathbb{E}_{\omega_{k+1}}(g(\omega_1, ..., \omega_k, \cdot))$$

Which means

$$\mathbb{E}_p(fg|\omega_1, ..., \omega_k) \geq \mathbb{E}_p(f|\omega_1, ..., \omega_k)\mathbb{E}_p(g|\omega_1, ..., \omega_k)$$

So applying the tower law of expectation

$$\begin{aligned}
\mathbb{E}_p(fg) &= \mathbb{E}_p(\mathbb{E}_p(fg|\omega_1, ..., \omega_k)) \\
&\geq \mathbb{E}_p(\mathbb{E}_p(f|\omega_1, ..., \omega_k)\mathbb{E}_p(g|\omega_1, ..., \omega_k)) \\
&\geq \mathbb{E}_p(\mathbb{E}_p(f|\omega_1, ..., \omega_k))\mathbb{E}_p(\mathbb{E}_p(g|\omega_1, ..., \omega_k))] \\
&= \mathbb{E}_p(f)\mathbb{E}_p(g)
\end{aligned}$$

So we have shown the result for all increasing functions which depend on a finite number of edges. Now to extend this to arbitrary bounded increasing functions we consider the martingale convergence theorem. For our bounded increasing functions f and g we define:

$$f_n := \mathbb{E}_p(f|\omega_1, ..., \omega_n)$$
$$g_n := \mathbb{E}_p(g|\omega_1, ..., \omega_n)$$

Then (f_n) and (g_n) form sub-martingales since f and g are increasing. Then since we assumed f and g are bounded we can apply the Martingale convergence Theorem to see that f_n converges to f as $n \to \infty$ and similarly g_n converges to g. We do not give a rigorous proof of this here but the reader may wish to seek further information on Doob's Martingale convergence Theorem from [8].

$\square$

We also note a useful rearrangement of Harris' Inequality which captures the idea that the events A and B are positively correlated.

Corollary 4.4. *If $A, B \subseteq \Omega$ are increasing events then*

$$\mathbb{P}_p(A|B) \geq \mathbb{P}_p(A)$$

Intuitively this makes sense, the probability A occurs increases when we know more edges are open, and knowing B has occurred tells us about open edges.

We now prove a useful inequality for events that occur disjointly, first we define what we mean for events to occur disjointly.

Definition 4.5 (Disjoint realisation of events)**.** Let A and B be events. For a configuration $\omega \in A$, there is some witness set $I \subset E$ for ω such that for any other configuration ω' which satisfies:

$$\forall e \in I, \, \omega_e = \omega'_e$$

Then $\omega' \in A$, in essence the states of edges in I tell us the event A has occurred in ω.
We say the events A and B occur disjointly in ω, denoted $A \circ B$, if there exists disjoint witnesses I and J for A and B respectively. I.e. $I \cap J = \emptyset$

This concept is different to the intersection of events, since intuitively the edges on which each event occurs cannot overlap. For example given some $n \times n$ box of $\mathbb{Z}^2$, the event there is an open crossing from top to bottom and the event there is an open crossing left to right cannot occur disjointly. Since the open paths will cross at some point. We now state the inequality we aim to prove.

Theorem 4.6 (BK inequality). *If $A, B \subseteq \Omega$ are increasing events then*

$$\mathbb{P}_p(A \circ B) \leq \mathbb{P}_p(A)\mathbb{P}_p(B)$$

Proof. We follow the proof given by [2] which is easier than the original. We work by induction on the size of the edge set E which we denote by n, remembering the sample space $\Omega = \{0,1\}^E$. When $n = 0$ the inequality holds trivially. Supposing as our inductive hypothesis that the inequality holds when the edge set has size less than n. We show it also holds when E has size n. First we define for any subset $D \subseteq \{0,1\}^E$:

$$D_0 = \{(\omega_1, ..., \omega_{n-1}) : (\omega_1, ..., \omega_{n-1}, 0) \in D\}$$
$$D_1 = \{(\omega_1, ..., \omega_{n-1}) : (\omega_1, ..., \omega_{n-1}, 1) \in D\}$$

Then we can decompose the probability of D since the states of edges are independent.

$$\mathbb{P}_p(D) = p\mathbb{P}_p(D_1) + q\mathbb{P}_p(D_0)$$

Now for our increasing $A, B \subseteq \{0,1\}^E$, define $C = A \circ B$, we now see that

$$C_0 = A_0 \circ B_0$$

$$C_1 = (A_1 \circ B_0) \cup (A_0 \circ B_1)$$

Applying our assumption that A and B are increasing its clear that $A_0 \subseteq A_1$ and $B_0 \subseteq B_1$. Thus we find

$$C_0 = A_0 \circ B_0 \subseteq (A_1 \circ B_0) \cap (A_0 \circ B_1)$$

$$C_1 = (A_1 \circ B_0) \cup (A_0 \circ B_1) \subseteq A_1 \circ B_1$$

Using this decomposition in conjunction with our inductive hypothesis we find

$$\mathbb{P}_p(C_0) = \mathbb{P}_p(A_0 \circ B_0) \leq \mathbb{P}_p(A_0)\mathbb{P}_p(B_0)$$
$$\mathbb{P}_p(C_1) \leq \mathbb{P}_p(A_1 \circ B_1) = \mathbb{P}_p(A_1)\mathbb{P}_p(B_1)$$

Now we can also get by adding the probabilities of C_0 and C_1 that

$$\mathbb{P}_p(C_0) + \mathbb{P}_p(C_1) \leq \mathbb{P}_p((A_1 \circ B_0) \cap (A_0 \circ B_1)) + \mathbb{P}_p((A_1 \circ B_0) \cup (A_0 \circ B_1))$$
$$= \mathbb{P}_p(A_0 \circ B_1) + \mathbb{P}_p(A_1 \circ B_0)$$
$$\leq \mathbb{P}_p(A_0)\mathbb{P}_p(B_1) + \mathbb{P}_p(A_1)\mathbb{P}_p(B_0)$$

The next step is to combine these by finding

$$(1-p)^2\mathbb{P}_p(C_0) + p^2\mathbb{P}_p(C_1) + p(1-p)(\mathbb{P}_p(C_0) + \mathbb{P}_p(C_1))$$

$$= ((1-p)^2 + p(1-p))\mathbb{P}_p(C_0) + (p^2 + p(1-p))\mathbb{P}_p(C_1) = (1-p)\mathbb{P}_p(C_0) + p\mathbb{P}_p(C_1)$$

Then looking at the right-hand sides

$$(1-p)\mathbb{P}_p(C_0) + p\mathbb{P}_p(C_1) \leq (1-p)^2[\mathbb{P}_p(A_0)\mathbb{P}_p(B_0)]$$
$$+ p^2[\mathbb{P}_p(A_1)\mathbb{P}_p(B_1)]$$
$$+ p(1-p)[\mathbb{P}_p(A_0)\mathbb{P}_p(B_1) + \mathbb{P}_p(A_1)\mathbb{P}_p(B_0)]$$

Which simplifies to

$$(1-p)\mathbb{P}_p(C_0) + p\mathbb{P}_p(C_1) \leq [p\mathbb{P}_p(A_1) + (1-p)\mathbb{P}_p(A_0)][p\mathbb{P}_p(B_1) + (1-p)\mathbb{P}_p(B_0)]$$

Then using $\mathbb{P}_p(D) = p\mathbb{P}_p(D_1) + (1-p)\mathbb{P}_p(D_0)$ we see that

$$\mathbb{P}_p(A \circ B) = \mathbb{P}_p(C) \leq \mathbb{P}_p(A)\mathbb{P}_p(B)$$

$\square$

The restriction of the BK inequality to increasing events was removed by Reimer, but it is not particularly useful to us so we do not prove it.

Using the BK inequality we can now show a useful inequality more specific to percolation.

Corollary 4.7. *For a finite subset $S \subset \mathbb{Z}^d$ which contains the origin, and some $x \notin S$,*

$$\mathbb{P}_p(0 \longleftrightarrow x) \sum_{y \in \partial S} \mathbb{P}_p(0 \xleftrightarrow{S} y)\mathbb{P}_p(y \longleftrightarrow x)$$

Where $0 \xleftrightarrow{S} y$ occurs if 0 is connected to y by open edges in S. If no set S is given we allow any open edges. Also, ∂S denotes the vertex boundary of S, i.e. $\forall s \in \partial S$, the neighbourhood of s contains some $x \in \mathbb{Z}^d \setminus S$

Proof. todo Work on an n by n box $\square$

5 Conformal Invariance

Discuss what conformal invariance is and perhaps talk about the result on the triangular lattice due to Smirnov

6 Simulation

We discuss various ways of exploring percolation via simulations.

6.1 The "water level" of open crossings in finite boxes

Discuss how uniform R.V.s can be assigned to edges to simulate percolation. Then plot the threshold for open crossings across some size box. Then see the sharpening of said plot when the box size increases.

As we have seen previously the task of determining the percolation threshold for a given lattice structure is highly non-trivial
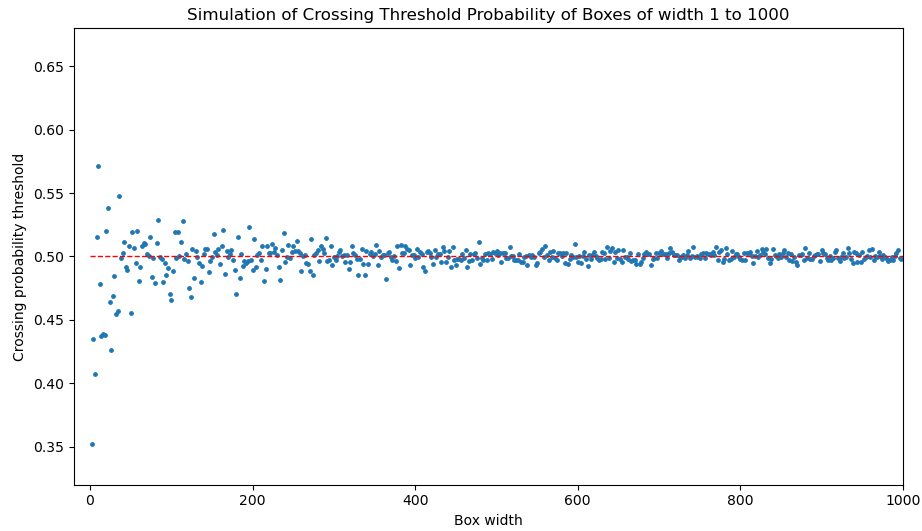
Figure 1: Crossing thresholds for simulated lattices

6.2 Finding the pivotal edges in a finite box

Discuss an algorithm to find the pivotal edges in the percolation box, perhaps relate to Russo's formula for pivotal edges.

6.3 Fast Monte Carlo algorithm for percolation

Discuss the Newman and Ziff paper "A fast Monte Carlo algorithm for site or bond percolation". Explaining the data structures etc, and perhaps implement it.

7 Conclusion

Is a conclusion necessary for this project?

References

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