

Percolation in the k -nearest neighbor graph

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Abstract

Let \mathcal{P} be a Poisson process of intensity one in \mathbb{R}^2 . For a fixed integer k , join every point of \mathcal{P} to its k nearest neighbors, creating a directed random geometric graph $\vec{G}_k(\mathbb{R}^2)$. We prove bounds on the values of k that, almost surely, result in an infinite connected component in $\vec{G}_k(\mathbb{R}^2)$ for various definitions of “component”. We also give high confidence results for the exact values of k needed. In particular, for percolation on the underlying (undirected) graph of $\vec{G}_k(\mathbb{R}^2)$, we prove that $k = 11$ is sufficient, and show with high confidence that $k = 3$ is the actual threshold for percolation.

1 Introduction

Let \mathcal{P} be a Poisson process of intensity one in \mathbb{R}^d , $d \geq 2$. For a fixed integer k , we join every point of \mathcal{P} to its k nearest neighbors, creating a directed random geometric graph $\vec{G}_k(\mathbb{R}^d)$ in which every vertex has out-degree exactly k . In this paper we shall mainly consider the case $d = 2$. The connectivity of these graphs restricted to a finite region in \mathbb{R}^2 was studied in [13, 2, 3]. Here we shall study percolation in the infinite region \mathbb{R}^2 , i.e., the existence or otherwise of infinite connected graph components. Since we are dealing with directed graphs, there are several possible definitions we can use for percolation.

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- U:** The underlying undirected graph has an infinite component.
- O:** The directed graph has an infinite directed out-component.
- I:** The directed graph has an infinite directed in-component.
- S:** The directed graph has an infinite strongly connected component.
- B:** The directed graph has an infinite component consisting of bidirectional edges.

Here an *out-component* is a subgraph with a spanning subtree whose edges are all directed away from a root vertex, while an *in-component* is a subgraph with a spanning subtree whose edges are all directed towards a root vertex. As all degrees are almost surely finite in this model, conditions **O** and **I** are equivalent to the existence of infinite paths directed away, respectively towards, a root vertex. A *strongly connected* subgraph is one where there are directed paths from u to v for any choice of vertices u and v in the component. An edge uv is *bidirectional* if both \vec{uv} and \vec{vu} lie in $\vec{G}_k(\mathbb{R}^d)$. Clearly we have the following implications

$$\mathbf{B} \Rightarrow \mathbf{S} \Rightarrow (\mathbf{I} \text{ and } \mathbf{O}), \quad (\mathbf{I} \text{ or } \mathbf{O}) \Rightarrow \mathbf{U}.$$

From now on, let \mathbf{X} denote any of **U**, **O**, **I**, **S**, or **B**. Let $\theta_{\mathbf{X}}(k, d)$ denote the probability that $G_k(\mathbb{R}^d)$ contains an infinite connected component according to definition **X**.

Lemma 1. *For all values of k , d , and X , $\theta_{\mathbf{X}}(k, d) \in \{0, 1\}$.*

Proof. Let E be the event that $\vec{G}_k(\mathbb{R}^d)$ has an infinite \mathbf{X} -component. By Kolmogorov's 0-1 law, it is enough to show that E is a *tail event*, i.e., it depends only on the vertices at distance $> K$ from the origin for any value of K . Fix $K > 0$. Then for any $\varepsilon > 0$ there is a $K_\varepsilon > K$ such that the probability that there exists a vertex at distance at least K_ε from the origin that is joined to some vertex within K of the origin is less than ε . Indeed, one can estimate the expected number of vertices v at distance at least L from 0 whose k th nearest neighbor is at distance more than $d(v, 0) - K$ as

$$\int_L^\infty \sum_{i=0}^{k-1} e^{-V_d(r-K)^d} \frac{(V_d(r-K)^d)^i}{i!} S_d r^{d-1} dr$$

where S_d and V_d are the surface area and volume respectively of a unit d dimensional ball. The sum in the integral above is a polynomial times a (super-) exponentially decreasing function, so the integral converges. Hence the integral can be made arbitrarily small by suitable choice of L .

Now there are almost surely only finitely many vertices within distance K_ε of the origin, so up to probability zero events, E is also the event that there is an infinite component in $G_k(\mathbb{R}^d) \setminus B(0, K_\varepsilon)$. (For each choice of \mathbf{X} , \mathbf{X} -percolation is unaffected by the removal of a finite number of vertices.) But with probability $1 - \varepsilon$ this does not depend on the choice of points within distance K of the origin. Since this holds for all $\varepsilon > 0$, E is, up to a set of probability zero, equal to an event that does not depend on points within distance K of the origin. Since this is true for all $K \in \mathbb{N}$, say, E is, up to a probability zero event, a tail event. Thus $\theta_{\mathbf{X}}(k, d) = \mathbb{P}(E) \in \{0, 1\}$. \square

It is clear that $\theta_{\mathbf{X}}(k, d)$ is non-decreasing in k . Define $k_{\mathbf{X},d}$ to be the *critical out-degree*, i.e., the smallest k such that $\theta_{\mathbf{X}}(k, d) > 0$ (equivalently $\theta_{\mathbf{X}}(k, d) = 1$). Our aim in this paper is to present rigorous bounds on the critical out-degrees $k_{\mathbf{X},2}$ for each choice of \mathbf{X} described above (Section 2, Theorem 2), as well as providing high confidence results for their exact values (Section 3).

2 Bounds

It has been shown by Häggström and Meester [7] that $k_{\mathbf{U},d} > 1$ for all d (see also [12]) and that $k_{\mathbf{U},d} = 2$ for sufficiently large d . (Actually, the proof in [7] shows that $k_{\mathbf{O},d} = 2$ for sufficiently large d .) Also, Teng and Yao [12] have shown that $k_{\mathbf{U},2} \leq 213$. We improve and generalize this last bound as follows.

Theorem 2. $k_{\mathbf{U},2} \leq 11$, $k_{\mathbf{O},2}, k_{\mathbf{I},2}, k_{\mathbf{S},2} \leq 13$, and $k_{\mathbf{B},2} \leq 15$.

To prove this result we shall compare the process to various bond percolation models on \mathbb{Z}^2 . In these models, the states of the edges will not be independent, however they will satisfy the following property.

Definition 1. A bond percolation model is 1-independedent if whenever E_1 and E_2 are sets of edges at graph distance at least 1 from each another (i.e., if no edge of E_1 is incident to an edge of E_2) then the state of the edges in E_1 is independent of the state of the edges in E_2 .

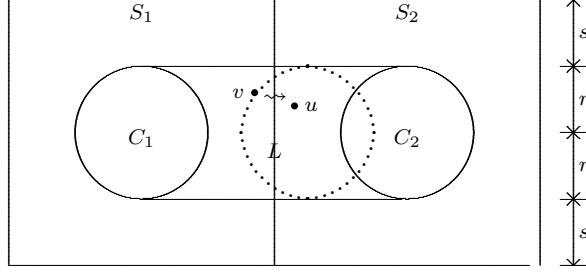


Figure 1: The regions defining $\mathcal{E}_{\mathbf{X},S_1,S_2}$, $\mathcal{E}'_{\mathbf{X},S_1,S_2}$, and the rolling ball D_v (dotted circle).

We shall use the following result, which is Theorem 2 of [4] (together with the remarks following its proof).

Theorem 3. *If every edge in a 1-independent bond percolation model on \mathbb{Z}^2 is open with probability at least 0.8639, then almost surely there is an infinite open component. Moreover, for any bounded region, there is almost surely an cycle of open edges surrounding this region.* \square

Proof of Theorem 2. Let us first consider the case of \mathbf{U} -percolation. Write $u \rightsquigarrow_{\mathbf{U}} v$ (or just $u \rightsquigarrow v$) if either $u\vec{v} \in \vec{G}_k(\mathbb{R}^2)$ or $v\vec{u} \in \vec{G}_k(\mathbb{R}^2)$, i.e., if uv is an edge of the underlying undirected graph. (Of course, this definition is symmetric, so that $u \rightsquigarrow v$ iff $v \rightsquigarrow u$. However, when we generalize this argument to the other types of percolation this may no longer hold.)

For percolation we need to find an infinite \rightsquigarrow -path, i.e., a sequence u_1, u_2, \dots with $u_i \rightsquigarrow u_{i+1}$ for all i . Consider the rectangular region consisting of two adjacent squares S_1, S_2 shown in Figure 1. Both S_1 and S_2 have side length $2r + 2s$, where r and s are to be chosen later. Also, S_2 may be to the right, left, above or below S_1 , in which case Figure 1 should be rotated accordingly. We define the *basic good event* $\mathcal{E}_{\mathbf{U},S_1,S_2}$ to be the event that every vertex u_1 in the central disk C_1 of S_1 is joined to at least one vertex v in the central disk C_2 of S_2 by a \rightsquigarrow -path, regardless of the state of the Poisson process outside of $S_1 \cup S_2$, and moreover that C_1 contains at least one vertex.

Now consider the following percolation model on \mathbb{Z}^2 . Each vertex $(i, j) \in \mathbb{Z}^2$ corresponds to a square $[Ri, R(i+1)] \times [Rj, R(j+1)]$ in \mathbb{R}^2 , where $R = 2r + 2s$, and an edge is open between adjacent vertices (corresponding to squares S_1 and S_2) if *both* the corresponding basic good events $\mathcal{E}_{\mathbf{U},S_1,S_2}$ and $\mathcal{E}_{\mathbf{U},S_2,S_1}$ hold. Note

that this is indeed a 1-independent model on \mathbb{Z}^2 since the event $\mathcal{E}_{\mathbf{U}, S_1, S_2}$ depends only on the Poisson process within the region $S_1 \cup S_2$, and thus sets of edges at distance at least one apart in \mathbb{Z}^2 depend on the Poisson process in disjoint regions of \mathbb{R}^2 . Any open path p_1, p_2, p_3, \dots in \mathbb{Z}^2 , corresponds to a sequence of basic good events $\mathcal{E}_{S_1, S_2}, \mathcal{E}_{S_2, S_3}, \dots$ that occur, where S_i is the square associated to p_i . Every vertex u_1 of the original Poisson process that lies in the central disk C_1 of S_1 now has an infinite \rightsquigarrow -path leading away from it, since one can find points u_i in the central disk of S_i and \rightsquigarrow -paths from u_{i-1} to u_i inductively for every $i > 1$. In particular, each such u_1 lies in an infinite \mathbf{U} -component. Moreover, such vertices exist in C_1 , so there is an infinite \mathbf{U} -component. From the bounds in Table 1 (which will be proved in Lemma 4 below), we see that for $k = 11$ one can choose r and s so that

$$\mathbb{P}(\mathcal{E}_{\mathbf{U}, S_1, S_2} \text{ fails}) \leq 0.0653 < 0.06805 = (1 - 0.8639)/2,$$

so in particular

$$\mathbb{P}(\mathcal{E}_{\mathbf{U}, S_1, S_2} \text{ and } \mathcal{E}_{\mathbf{U}, S_2, S_1} \text{ hold}) \geq 0.8639.$$

The result now follows from Theorem 3.

For $k_{\mathbf{B}, 2}$ we follow the same proof as above, except that we define $u \rightsquigarrow_{\mathbf{B}} v$ to hold if *both* $u\vec{v} \in \vec{G}_k(\mathbb{R}^2)$ and $v\vec{u} \in \vec{G}_k(\mathbb{R}^2)$. The event $\mathcal{E}_{\mathbf{B}, r, s}$ is now defined as for $\mathcal{E}_{\mathbf{U}, r, s}$ except that we use $\rightsquigarrow_{\mathbf{B}}$ in place of $\rightsquigarrow_{\mathbf{U}}$, and the result follows from the bound (for $k = 15$) given in Table 1, since $0.0676 < 0.06805$.

Similarly, the bounds in Table 1 give $k_{\mathbf{O}, 2} \leq 13$, where we follow the same proof as above using $u \rightsquigarrow_{\mathbf{O}} v$, which is defined to hold if $u\vec{v} \in \vec{G}_k(\mathbb{R}^2)$. In this case \rightsquigarrow is not symmetric, however the above proof still gives an infinite outwards directed path from some vertex.

At first sight it seems from Table 1 that the bound for $k_{\mathbf{I}, 2}$ (using $u \rightsquigarrow_{\mathbf{I}} v$, which holds iff $v\vec{u} \in \vec{G}_k(\mathbb{R}^2)$) will not be as good. Moreover, we do not have an analogous proof for $k_{\mathbf{S}, 2}$. However, it turns out that our bound on $k_{\mathbf{O}, 2}$ applies to $k_{\mathbf{S}, 2}$ as well. To see this, note that the above argument shows that (for $k = 13$, suitable r , s , and $\rightsquigarrow_{\mathbf{O}}$) we have an infinite path $p_1 p_2 \dots$ in \mathbb{Z}^2 corresponding to a sequence of squares S_1, S_2, \dots , with each edge $p_i p_{i+1}$ corresponding to basic good events $\mathcal{E}_{\mathbf{O}, S_i, S_{i+1}}, \mathcal{E}_{\mathbf{O}, S_{i+1}, S_i}$ in *both* directions. Let Z_i be the set of vertices in $\vec{G}_k(\mathbb{R}^2)$ that are in the central disk of S_i . Then Z_1 is almost surely finite. Fix $N > 0$. Each $v_{i_1} \in Z_1$ is joined to some $u_{i_1} \in Z_N$ by a directed path in $\vec{G}_k(\mathbb{R}^2)$. But similarly u_{i_1} is joined by a directed path to some element $v_{i_2} \in Z_1$. Iterating this

Table 1: Upper bounds on $\min_{r,s} \mathbb{P}(\mathcal{E}_{\mathbf{X},r,s} \text{ fails})$. (All numbers rounded up.)

$\mathbf{X} \backslash k$	9	10	11	12	13	14	15
U	.1786	.1090	.0653	.0386	.0225	.0130	.0075
O	.3424	.2215	.1402	.0871	.0533	.0322	.0192
I	.4906	.3511	.2476	.1725	.1189	.0812	.0551
B	.6217	.4472	.3151	.2183	.1492	.1009	.0676

process starting at v_{i_2}, v_{i_3}, \dots in turn we must eventually repeat some vertex of Z_1 . Hence some vertex of Z_1 lies on a directed closed trail meeting at least N vertices (at least one from each of Z_1, Z_2, \dots, Z_N , which are disjoint sets). Since this holds for every N , and Z_1 is finite, there exists a single $v \in Z_1$ which lies on arbitrarily large directed closed trails. Thus in particular v lies in an infinite strong component. Thus $k_{\mathbf{S},2} \leq 13$. Finally **S**-percolation implies **I**-percolation, so $k_{\mathbf{I},2} \leq 13$ also holds. \square

We note that in the above proof we declared an edge in \mathbb{Z}^2 to be open if *both* $\mathcal{E}_{\mathbf{U},S_1,S_2}$ and $\mathcal{E}_{\mathbf{U},S_2,S_1}$ held. It would seem that (at least in the **U** and **B** cases) that we would need only one, say $\mathcal{E}_{\mathbf{U},S_1,S_2}$ with S_2 either to the right or above S_1 . However, in this case one needs an *oriented* open path in \mathbb{Z}^2 , which at each step goes either to the right or up, to obtain an infinite $\rightsquigarrow_{\mathbf{U}}$ -path. This is because $\mathcal{E}_{\mathbf{U},S_1,S_2}$ and $\mathcal{E}_{\mathbf{U},S_3,S_2}$ do not force a path from S_1 to S_3 . Unfortunately no good bounds appear to have been proved for 1-independent oriented bond percolation in \mathbb{Z}^2 , and in any case such bounds are unlikely to improve much on the method used above.

To complete the proof of Theorem 2, we need to show the following.

Lemma 4. *The probabilities that the $\mathcal{E}_{\mathbf{X},r,s}$ fail can be bounded (for suitable choices of r and s) by the values given in Table 1.*

Proof. To bound the probability that a basic good event fails, we shall use the following “rolling ball” method. Let C_1 , C_2 , and L be as in Figure 1. (L is the region between the two disks C_1 and C_2 .) For $\mathbf{X} \in \{\mathbf{U}, \mathbf{O}, \mathbf{I}, \mathbf{B}\}$, define $\mathcal{E}'_{\mathbf{X},S_1,S_2}$ to be the event that for every point $v \in C_1 \cup L$, there is a u such that

- (a) $v \rightsquigarrow_{\mathbf{X}} u$;

- (b) $\|u - v\| \leq s$; and
- (c) $u \in D_v$, where D_v is the disk of radius r inside $C_1 \cup L \cup C_2$ with v on its C_1 -side boundary (the dotted disk in Figure 1).

Note in particular that (b) implies that the condition $u \rightsquigarrow v$ in (a) is independent of the Poisson process outside of $S_1 \cup S_2$. This is because both u and v are at distance at least s from the exterior of $S_1 \cup S_2$, so the event that u is among the k nearest neighbors of v , or that v is among the k nearest neighbors of u , only depends on the points within $S_1 \cup S_2$. If $\mathcal{E}'_{\mathbf{X}, S_1, S_2}$ holds, then every vertex v in C_1 must be joined by a \rightsquigarrow -path to a vertex in C_2 , since each vertex in $C_1 \cup L$ is joined to a vertex whose disk D_v is further along in $C_1 \cup L \cup C_2$. Thus if we let \mathcal{F}_{S_1} be the event that there is at least one vertex in C_1 , we have $\mathcal{E}'_{\mathbf{X}, S_1, S_2} \cap \mathcal{F}_{S_1} \subseteq \mathcal{E}_{\mathbf{X}, S_1, S_2}$. The probability that \mathcal{F}_{S_1} fails is simply the probability that there is no vertex in S_1 , which is $e^{-\pi r^2}$. The probability that $\mathcal{E}'_{\mathbf{X}, S_1, S_2}$ fails is bounded by the expected number of points u for which the above conditions (a)–(c) fail. The expected number of points in $C_1 \cup L$ is $|C_1 \cup L| = 2r(2r + 2s)$. Thus

$$\mathbb{P}(\mathcal{E}'_{\mathbf{X}, S_1, S_2} \text{ fails}) \leq 2r(2r + 2s)p_{\mathbf{X}, r, s} \quad (1)$$

where $p_{\mathbf{X}, r, s}$ is the probability that (a)–(c) fail for some fixed v . Note that this probability is independent of the location of v in $C_1 \cup L$.

To bound $p_{\mathbf{X}, r, s}$ we consider the probability that the vertex u *closest* to v inside D_v fails (a)–(c) (or does not exist). Let us consider the $\mathbf{X} = \mathbf{U}$ case first. Condition on the existence of a vertex $u \in \mathcal{P} \cap D_v$, and define the regions A , B , and C as in Figure 2. Let $p_{\mathbf{U}}(u)$ be the probability that u is the closest point to v inside D_v , but that $v \rightsquigarrow_{\mathbf{U}} u$ fails. Then there are at least k points of the Poisson process in $A = B(v, \alpha) \setminus D_v$, at least k points in $C = B(u, \alpha)$, but no points in $B = B(v, \alpha) \cap D_v$. Thus

$$\begin{aligned} p_{\mathbf{U}}(u) &= \sum_{i=0}^{\infty} \sum_{j, l \geq \max\{0, k-i\}} \mathbb{P}(\text{Po}(|A \cap C|) = i) \mathbb{P}(\text{Po}(|A \setminus C|) = j) \\ &\quad \times \mathbb{P}(\text{Po}(|C \setminus (A \cup B)|) = l) \mathbb{P}(\text{Po}(|B|) = 0) \\ &= \sum_{i=0}^{\infty} \sum_{j, l \geq \max\{0, k-i\}} \frac{|A \cap C|^i |A \setminus C|^j |C \setminus (A \cup B)|^l}{i! j! l!} e^{-|A \cup B \cup C|}, \end{aligned} \quad (2)$$

since if there are i points in $A \cap C$ then there are at least $k - i$ in $A \setminus C$ and $C \setminus (A \cup B)$, and none in B . Now the probability that there is no u satisfying

conditions (a)–(c) above is bounded by

$$p_{\mathbf{U},r,s} \leq e^{-|D_v \cap B(v,s)|} + \int_{u \in D_v \cap B(v,s)} p_{\mathbf{U}}(u) du. \quad (3)$$

The first term being the probability that there is no u satisfying (b) and (c), and the integral gives the probability that such a u exists, but that the closest one to v fails (a). Explicit calculation of this upper bound to $p_{\mathbf{U},r,s}$ is rather unpleasant due to the calculation of the areas above. However, numerical bounds can be computed (see Appendix A and [1]). Finally,

$$\mathbb{P}(\mathcal{E}_{\mathbf{U},S_1,S_2} \text{ fails}) \leq \mathbb{P}(\mathcal{F}_{S_1} \text{ fails}) + \mathbb{P}(\mathcal{E}'_{\mathbf{U},S_1,S_2} \text{ fails}) \leq e^{-\pi r^2} + 2r(2r+2s)p_{\mathbf{U},r,s}, \quad (4)$$

and this bound can be minimized over various values of r and s . The minimum values obtained are listed in Table 1 (row \mathbf{U}) for various values of k .

The calculation for the other cases is exactly analogous. For \mathbf{B} we replace (2) by

$$p_{\mathbf{B}}(u) = \sum_{i=0}^{\infty} \sum_{\max\{j,l\} \geq \max\{0,k-i\}} \frac{|A \cap C|^i |A \setminus C|^j |C \setminus (A \cup B)|^l}{i!j!l!} e^{-|A \cup B \cup C|}$$

since now we require *either* at least k points in A *or* at least k points in C for $v \rightsquigarrow u$ to fail. For \mathbf{O} , failure occurs when there are at least k points in A , so (2) becomes

$$p_{\mathbf{O}}(u) = \sum_{j=k}^{\infty} \frac{|A|^j}{j!} e^{-|A \cup B|}.$$

For completeness we also consider the case \mathbf{I} , where at least k points in C is required for $v \rightsquigarrow u$ to fail. In this case (2) becomes

$$p_{\mathbf{I}}(u) = \sum_{l=k}^{\infty} \frac{|C \setminus B|^l}{l!} e^{-|B \cup C|}.$$

In each case the bounds (3) and (4) generalize to

$$\mathbb{P}(\mathcal{E}_{\mathbf{X},S_1,S_2} \text{ fails}) \leq e^{-\pi r^2} + 2r(2r+2s) \left(e^{-|D_v \cap B(v,s)|} + \int_{u \in D_v \cap B(v,s)} p_{\mathbf{X}}(u) du \right), \quad (5)$$

and the minimum value (over r and s) found is listed in Table 1. \square

Let us remark, that we proved the weaker bound $k_{\mathbf{U},2} \leq 13$ already in 2003, and mentioned it in several conferences.

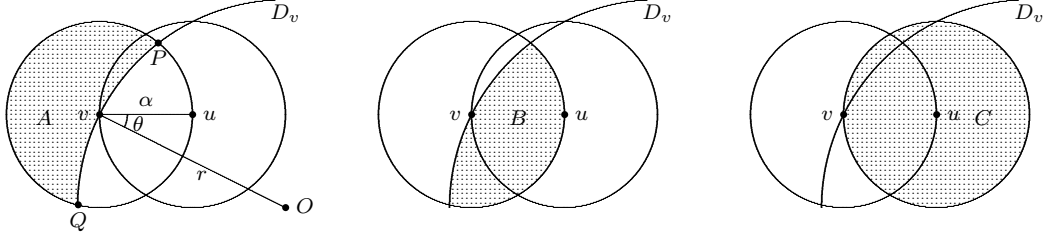


Figure 2: Areas $A = B(v, \alpha) \setminus D_v$, $B = B(v, \alpha) \cap D_v$, and $C = B(u, \alpha)$.

3 High confidence results

In this section, we evaluate the critical out-degrees $k_{\mathbf{X},2}$ with *high confidence*. Here, high confidence means that we reduce to showing that if a certain high dimensional integral exceeds a given value, then percolation occurs (respectively does not occur). Unfortunately, the integral is impractical to evaluate exactly, so it is estimated using a Monte-Carlo approach. The value obtained then gives a proof of the result, subject to the proviso that the random numbers used in the Monte-Carlo calculation did not lie in the very small region of the sample space that gives a misleading value for this integral. (See [6, 4] for examples of this approach being applied to other percolation questions.)

Results. *With high confidence,* $k_{\mathbf{U},2} = 3$, $k_{\mathbf{O},2} = k_{\mathbf{I},2} = k_{\mathbf{S},2} = 4$, $k_{\mathbf{B},2} = 5$.

To show percolation in the cases $\mathbf{X} \in \{\mathbf{U}, \mathbf{O}, \mathbf{B}\}$ (with $k = 3, 4, 5$ respectively), we choose r and s as above. We then generate a random instance of the process inside $S_1 \cup S_2$ and test for the following conditions:

- UB₁ For more than half of the vertices $v \in C_1$ there are \rightsquigarrow -paths from v to more than half the vertices of C_2 , regardless of the state of the process outside of $S_1 \cup S_2$.
- UB₂ Similarly, more than half of the vertices of C_2 have paths to more than half the vertices of C_1 , regardless of the state of the process outside of $S_1 \cup S_2$.

As before, it is clear that if we have a sequence of distinct squares S_1, S_2, \dots with the above holding in $S_i \cup S_{i+1}$ for all i , then there will be an infinite \rightsquigarrow -path from some vertex in C_1 . (The conditions UB₁ and UB₂ were chosen in place of $\mathcal{E}_{\mathbf{X}, S_1, S_2}$ and $\mathcal{E}_{\mathbf{X}, S_2, S_1}$ since in general they have a higher probability of success. Note that requiring strictly more than half the vertices of C_i to have a property implies that

at least one vertex must exist in C_i .) Also, UB_1 and UB_2 depend only on the Poisson process within $S_1 \cup S_2$. Hence by Theorem 3 we only need to show that these conditions hold with probability at least 0.8639. The condition that the path should be independent of the process outside of $S_1 \cup S_2$ is simply obtained by ignoring any edges \vec{uv} of $\vec{G}_k(S_1 \cup S_2)$ where $d(u, v) > d(u, \partial(S_1 \cup S_2))$, since only edges \vec{uv} with $d(u, v) \leq d(u, \partial(S_1 \cup S_2))$ are guaranteed to exist in $\vec{G}_k(\mathbb{R}^2)$.

Using a computer program we generated many instances, and counted the proportion of times these conditions held. The results are listed in the top half of Table 2. Using a similar argument as in the proof of Theorem 2, the infinite path in \mathbb{Z}^2 in the $\mathbf{X} = \mathbf{O}$ case actually gives us an infinite strong component, so in fact gives a bound for k_S . From these we calculate the confidence level, i.e., the probability p that these results (or better) could be obtained if the true probability of success was < 0.8639 . In all cases considered p is ludicrously small.

To show *lack* of percolation in the cases $\mathbf{X} \in \{\mathbf{U}, \mathbf{I}, \mathbf{O}, \mathbf{B}\}$ (with $k = 2, 3, 3, 4$ respectively), we generate, for suitable r, s , instances of the process in $S_1 \cup S_2$ and check the following condition holds:

LB₁ Regardless of the state outside $S_1 \cup S_2$, there is no \rightsquigarrow -path from outside of $S_1 \cup S_2$ that crosses the line segment that joins the center point of S_1 to the center point of S_2 (see Figure 3).

Once again we define a percolation model of \mathbb{Z}^2 by declaring an edge open if LB₁ holds in the corresponding rectangle $S_1 \cup S_2$. This model is also clearly 1-independent. Suppose the probability that LB₁ occurs is at least 0.8639. Then by Theorem 3 there are open cycles in the corresponding \mathbb{Z}^2 process surrounding any bounded region. If an infinite \rightsquigarrow -path existed starting in some such region, then it would have to cross this cycle, and in particular cross the central line segment in one of rectangles $S_1 \cup S_2$ corresponding to an open edge in this cycle. However, this would contradict condition LB₁ for this edge.

Note that we could have demanded in LB₁ only that there is no path from one boundary point to another boundary point that crosses the center line. However, the condition given is easier to test for, and is sufficient for our purposes.

To test whether an edge of a \rightsquigarrow -path could come from outside of $S_1 \cup S_2$ to a vertex $v \in S_1 \cup S_2$ is somewhat harder. In the $\mathbf{X} = \mathbf{I}$ case, one can just test whether or not the k nearest neighbors of v are all closer to v than the boundary.

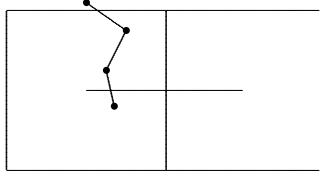


Figure 3: Condition LB_1 requires that there is no path from outside $S_1 \cup S_2$ crossing the line segment joining the centers of S_1 and S_2 .

Table 2: Results of Monte-Carlo simulation

Test	(r, s)	Successes	Trials	Confidence
$k_{\mathbf{U}} \leq 3$	$(18, 2)$	9984	10000	$p < 10^{-597}$
$k_{\mathbf{S}}(k_{\mathbf{O}}) \leq 4$	$(18, 2)$	9564	10000	$p < 10^{-208}$
$k_{\mathbf{B}} \leq 5$	$(18, 2)$	9960	10000	$p < 10^{-555}$
$k_{\mathbf{U}} > 2$	$(0, 60)$	9861	10000	$p < 10^{-430}$
$k_{\mathbf{O}} > 3$	$(0, 60)$	9667	10000	$p < 10^{-269}$
$k_{\mathbf{I}} > 3$	$(0, 60)$	9710	10000	$p < 10^{-299}$
$k_{\mathbf{B}} > 4$	$(0, 600)$	9460	10000	$p < 10^{-157}$

If not, we assume an edge (of $\vec{G}_k(\mathbb{R}^2)$) could leave $S_1 \cup S_2$, and so $u \rightsquigarrow v$ for some u outside of $S_1 \cup S_2$.

For the \mathbf{O} and \mathbf{U} cases, however, one must find the k nearest neighbors in $S_1 \cup S_2$ of *every* possible point outside of $S_1 \cup S_2$. It is easy to see that it is enough to check points that lie on the boundary of $S_1 \cup S_2$, however there are still an infinite number of these. Instead we use the following algorithm. Pick a point w on the boundary of $S_1 \cup S_2$ and find its $k + 2$ nearest neighbors in $S_1 \cup S_2$. Mark the $k + 1$ nearest neighbors of w as possibly having an edge from outside $S_1 \cup S_2$. Let d_i be the distance from w to its i th nearest neighbor in $S_1 \cup S_2$. Now advance by a distance $(d_{k+2} - d_k)/2$ along the boundary from w and then check this new point. Repeat this process until the entire boundary has been traversed. To see that this is sufficient, note that if $d(w', w) < (d_{k+2} - d_k)/2$, then the points that are not among the $k + 1$ nearest neighbors of w will all be further away from w' than the k nearest neighbors of w . Thus the k nearest neighbors (in $S_1 \cup S_2$) of w' will be a subset of the $k + 1$ nearest neighbors of w .

Finally, for the case \mathbf{B} we could use the above algorithm and also check edges

leaving $S_1 \cup S_2$ as in the **I** case. However, the above boundary searching algorithm is rather slow, and the size of the rectangle $S_1 \cup S_2$ needed was rather large in this case. Thus we have just checked edges leaving $S_1 \cup S_2$ as in the **I** case and assumed any $u \in S_1 \cup S_2$ with an edge leaving $S_1 \cup S_2$ also had an edge in from the exterior of $S_1 \cup S_2$. This makes the test LB_1 slightly more pessimistic, but in practice the difference in success rate was minimal, while the program ran much faster.

The results of these computer simulations are listed in the bottom half of Table 2. Once again, in all cases considered the result is shown with extremely high confidence. All our simulations used the alleged RC4 algorithm [11] for pseudo-random number generation. More details, including the **C** source code, can be found in [1].

A Calculation of the integral

To calculate the areas in Figure 2, write $\alpha = d(u, v)$ for the distance between u and v , and θ for the angle Ovu , where O is the center of D_v (see Figure 2). The following is a useful formula for the area $L(a, b, \phi)$ of a lune $D_a \setminus D_b$ consisting of the area inside a disk D_a of radius a and outside a disk D_b of radius b , and which makes an angle of $\phi \in (0, \pi)$ at each end (see Figure 4).

$$L(a, b, \phi) = \frac{a^2}{2}(2(\psi + \phi) - \sin 2(\psi + \phi)) - \frac{b^2}{2}(2\psi - \sin 2\psi)$$

where $\psi = \cos^{-1}\left(\frac{b^2 + c^2 - a^2}{2bc}\right) \in [0, \pi]$ and $c^2 = a^2 + b^2 - 2ab \cos \phi$. (The first term is the area above the line PQ inside of D_a , and the second is the area above PQ inside D_b .) It is useful also to define

$$L(a, b, \phi) = -L(b, a, -\phi) \quad \text{when } \phi < 0. \quad (6)$$

The angle of the lune A in Figure 2 is given by

$$\phi = \cos^{-1}(\alpha/2r) \in [0, \frac{\pi}{2}].$$

(The angle ϕ is also one of the angles at the base of the isosceles triangle OvP .) Clearly $\alpha \leq 2r$, and indeed, $\theta \in (-\phi, \phi)$, otherwise u would not lie in D_v . By symmetry we may assume that $\theta \geq 0$, so that $\theta \in [0, \phi)$. Now

$$|A \cup B| = |C| = \pi\alpha^2. \quad (7)$$

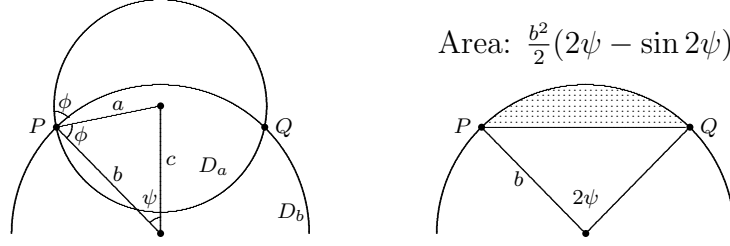


Figure 4: Lune used to define $L(a, b, \phi)$.

and a simple calculation shows that

$$|C \setminus (A \cup B)| = |(A \cup B) \setminus C| = \left(\frac{\pi}{3} + \frac{\sqrt{3}}{2}\right)\alpha^2. \quad (8)$$

Also, by the definition of $L(a, b, \phi)$,

$$|A| = L(\alpha, r, \phi). \quad (9)$$

We now calculate $|A \cap C|$. This calculation splits into three cases depending on whether the two intersection points of the boundary of D_v and $B(v, \alpha)$ (P and Q in Figure 2) lie inside of $B(u, \alpha)$. The result is

$$|A \cap C| = \begin{cases} L(\alpha, r, \theta) & \theta \leq \phi - \frac{\pi}{3}, \theta > \frac{\pi}{3} - \phi; \\ \frac{\alpha^2}{2}(\theta - \phi + \frac{\pi}{3}) + L(\alpha, r, \phi - \frac{\pi}{3}) & \theta > \phi - \frac{\pi}{3}, \theta > \frac{\pi}{3} - \phi; \\ |A| - |(A \cup B) \setminus C| + L(r, \alpha, \theta) & \theta \leq \frac{\pi}{3} - \phi. \end{cases} \quad (10)$$

(First case when $P, Q \notin B(u, \alpha)$, third case $P, Q \in B(u, \alpha)$, second case when $P \in B(u, \alpha)$, $Q \notin B(u, \alpha)$. Also, in the second case $\phi - \frac{\pi}{3}$ will be negative if $r < \alpha$, so we also use (6).) Combining (7)–(10) allows us to evaluate the areas necessary for the calculation of $p_{\mathbf{X}}(u)$.

To prove a bound on the integral in (5), we note that (10) is monotonic in θ , while (7)–(9) are independent of θ . One can check that the formulae giving $p_{\mathbf{X}}(u)e^{|B|}$ (as a function of α and θ) are monotonically increasing with both α and θ , for each choice of \mathbf{X} . Also $e^{-|B|}$ is decreasing in α . One can effectively bound $p_{\mathbf{X}}(u)$ over a small region R in the (α, θ) -plane by multiplying the maximum value of $p_{\mathbf{X}}(u)e^{|B|}$ in R by the maximum value of $e^{-|B|}$ over R . A bound on the total integral is then obtained by summing the bounds on the integrals over a suitable partition of $D_v \cap B(v, s)$ (see [1] for more details). Table 1 gives the results we obtained using this approach.

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