THE SIZE OF A POND IN 2D INVASION PERCOLATION

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Abstract

We consider invasion percolation on the square lattice. In [3] it has been proved that the probability that the radius of a so-called pond is larger than n, differs at most a factor of order $\log n$ from the probability that in critical Bernoulli percolation the radius of an open cluster is larger than n. We show that these two probabilities are, in fact, of the same order. Moreover, we prove an analogous result for the volume of a pond.

1   Introduction

Invasion percolation is a stochastic growth model of an interesting self-organised critical nature: it has characteristics that resemble critical Bernoulli percolation, even though the definition of the invasion process does not involve any parameter (see [4, 14]). Comparison of the two processes helps to gain new insights into both of them (see e.g. [5, 7, 10, 1]).

Recently a new comparison result, relating a so-called 'pond' in invasion percolation to a critical Bernoulli percolation cluster, has been proved in [3]. This result is sharpened and extended in the present paper.

In the remainder of this Section we define the invasion percolation model and state our main results. The proofs, and important prerequisites, are given in Section 2.

For general background on percolation, see [6].

Consider the hypercubic lattice $\mathbb{Z}^d$ with its set of nearest neighbour bonds $\mathbb{E}^d$. If an edge e has endpoints v and w, we write $e = \langle v, w \rangle$. For an arbitrary subgraph $G = (V, E)$ of $(\mathbb{Z}^d, \mathbb{E}^d)$, we define the outer boundary ΔG as

$$\Delta G = \{e = \langle v, w \rangle \in \mathbb{E}^d \ : \ e \notin E, \text{ but } v \in V \text{ or } w \in V\}.$$

Invasion percolation is defined as follows. Let $\tau(e), e \in \mathbb{E}^d$, be independent random variables, uniformly distributed on the interval $[0, 1]$. Using these variables, we construct inductively an increasing sequence $G_0, G_1, G_2, \dots$ of connected subgraphs of the lattice. G_0 only contains the origin. If $G_i = (V_i, E_i)$ has already been defined, we select the bond e_{i+1} which minimizes τ on ΔG_i, take $E_{i+1} = E_i \cup \{e_{i+1}\}$, and let G_{i+1} be the graph induced by the edge set E_{i+1}. The graph G_i is called the invaded cluster at time i, and $G_\infty = (V_\infty, E_\infty) = \cup_{i=0}^\infty G_i$ is the invaded region at time infinity.

   Invasion percolation can be coupled to Bernoulli bond percolation in the following standard way. Let $0 \le p \le 1$. For each bond e we say that e is p-open, if $\tau(e) < p$. One can then define, in an obvious way, p-open paths and p-open clusters, and the study of these objects corresponds with Bernoulli bond percolation with parameter p.

   From now on we will only consider the case when $d = 2$. It is well-known and easy to see that for every $p \in [0, 1]$ the following holds: once the invasion reaches an infinite p-open cluster, it never leaves it again. Further, it is a classical result for $2D$ Bernoulli percolation that for every $p > p_c$ (which equals $1/2$ on the square lattice) there is (a.s.) a p-open circuit that surrounds O and belongs to the infinite p-open cluster, and that (a.s.) there is no infinite p_c-open cluster. These properties easily imply that (a.s.) $\hat{\tau} := \max_{e \in E_\infty} \tau(e)$ exists and is larger than p_c. Let $\hat{e}$ denote the edge where the maximum is taken and suppose that it is added to the invasion cluster at step $\hat{i} + 1$. Following the terminology in [13], the graph $G_{\hat{i}} = (\hat{V}, \hat{E})$ is called a 'pond', or, more precisely, the first pond of O. Since the invasion can be started at any vertex v, not necessarily O, we have the more general notion 'first pond of v'.

   The above defined 'pond' is a very natural object (see [13] and [3]), and has several interpretations, for instance the following. In this (somewhat informal) interpretation each vertex $(x, y) \in \mathbb{Z}^2$ represents a 'polder': the square piece of (flat) land $(x-1/2, x+1/2) \times (y-1/2, y+1/2)$, surrounded by four dikes, corresponding with (the dual edges of) the four edges of (x, y). The heights of the dikes are the τ values of the corresponding edges. Now suppose that water is supplied from some external source to the polder represented by the vertex O. The water in this polder will rise until its level reaches the height of the lowest of its four dikes, say a. Then the water starts spilling over that dike, so that the level in the neighbouring polder (on the other side of the dike) starts to rise. If each of the other three dikes of that neighbouring polder is higher than a, the water in the polder of O will remain at level a until the above mentioned neighboring polder has reached this same water level, after which the level in both polders rises ('simultaneously') until it reaches the height of the lowest of the six dikes bounding the union of these two polders, etc. On the other hand, if the neighboring polder of O has a dike with height $b < a$, the water level in this polder will rise up to level b and then starts spilling over that dike to a third polder (while the level in the polder of O is still a) etc. In any case, (a.s.) eventually the water level at O will remain constant forever, namely at the level $\hat{\tau}$ defined above, and the 'connected' set of polders with the same final level as O is the above defined 'first pond'. Since water keeps being supplied, the 'surplus' water will spill over the lowest dike (corresponding with the earlier defined $\hat{e}$), on the boundary of this pond: the outlet from this first pond to a second (lower level) pond. For clarity we note that for each vertex in the latter pond, this pond plays the role of 'first pond'.

For further clarity we also note that the above 'hydrologic' interpretation has a more 'symmetric' version as follows: Now at *each* vertex there is an external water source (rain, e.g.). Again each polder has a final water level, and the maximal connected set of polders with the same final water level, containing a given vertex v, is the earlier defined (first) pond of v. Then, if $\hat{V}(v)$ denotes the vertices of the first pond of v, the collection $\{\hat{V}(v)\}_{v\in\mathbb{Z}^2}$ is a random partition of $\mathbb{Z}^2$ which is stationary under translations.

Before stating the results, we first fix some notation. Let $\hat{R} := \max\{|x| + |y| : (x,y) \in \hat{V}\}$ be the radius of the first pond. Let P_{cr} denote the product measure corresponding to critical Bernoulli bond percolation. Let $B(n)$ denote the box $[-n,n]^2$ and $\partial B(n) := B(n) \setminus B(n-1)$. Let A and B be sets of vertices. In the context of Bernoulli percolation, we denote the event that there is an open path from A to B by $\{A \leftrightarrow B\}$. In the context of invasion percolation we denote the event that there is a p-open path from A to B by $\{A \overset{p}{\leftrightarrow} B\}$. To indicate that there is an infinite open (or p-open) path from A, we use the same notation with B replaced by ∞.

We use the notation $g(n) \approx h(n)$, $n \to \infty$ to indicate that

$$\frac{\log g(n)}{\log h(n)} \to 1, \quad \text{as } n \to \infty,$$

and $g(n) \asymp h(n)$ to indicate that $g(n)/h(n)$ is bounded away from 0 an ∞.

Van den Berg, Peres, Sidoravicius and Vares have proved the following theorem:

Theorem. *[3, Proposition 1.3]*

$$P(\hat{R} \geq n) \approx P_{cr}(0 \leftrightarrow \partial B(n)), \ n \to \infty. \tag{1.1}$$

Using ideas and techniques from [7], we obtain the following improvement of the theorem above.

Theorem 1.

$$P(\hat{R} \geq n) \asymp P_{cr}(0 \leftrightarrow \partial B(n)). \tag{1.2}$$

Moreover, we show that not only the radius but also the volume of the pond behaves like that of a critical percolation cluster: Let

$$s(n) = n^2 P_{cr}(0 \leftrightarrow \partial B(n)) \qquad \mathcal{C}(0) = \{v \in \mathbb{Z}^2 : 0 \leftrightarrow v\}.$$

Theorem 2. *There exist constants $0 < c, c' < \infty$, such that*

$$cP_{cr}(0 \leftrightarrow \partial B(n)) \leq P_{cr}(|\mathcal{C}(0)| > s(n)) \leq P(|\hat{V}| > s(n))$$
$$\leq c' P_{cr}(0 \leftrightarrow \partial B(n)). \tag{1.3}$$

Corollary 3.

$$P(|\hat{V}| \geq n) \asymp P_{cr}(|\mathcal{C}(0)| \geq n). \tag{1.4}$$

Remark: These results, and the proofs in Section 2 also hold (with some obvious adaptations) for the triangular and the hexagonal lattice.

2 Proofs of the main results

In the following all the constants are strictly positive and finite without further mentioning.

2.1   Preliminaries

Let

$$\sigma(n, m, p) = P(\text{there is a } p\text{-open horizontal crossing of } [0, n] \times [0, m]),$$

where it is assumed that the crossing does not use bonds lying on the top or the bottom sides of the rectangle. Given $\varepsilon > 0$, we define, for $p > p_c$,

$$L(p, \varepsilon) = \min\{n : \sigma(n, n, p) \geq 1 - \varepsilon\}.$$

It is shown in [9, (1.24)], that there exists an $\varepsilon_0 > 0$ such that for all $\varepsilon \leq \varepsilon_0$, the scaling of $L(p, \varepsilon)$ is independent of ε in the sense that for all fixed $0 < \varepsilon_1, \varepsilon_2 \leq \varepsilon_0$ the ratio $L(p, \varepsilon_1)/L(p, \varepsilon_2)$ is bounded away from both 0 and ∞ as $p \downarrow p_c$. We let $L(p) = L(p, \varepsilon_0)$ for the entire proof. Below we list some properties of $L(p)$ that will play a crucial role in the proof of our results. The first two follow fairly easily from the definitions and standard arguments (see Section 2.2 in [7] for further explanation and references). The third is (a consequence of) a deep result in [9]

1. $L(p)$ is decreasing, right continuous and $L(p) \to \infty$ as $p \downarrow p_c$.

2. There is a constant D such that

$$\lim_{\delta \downarrow 0} \frac{L(p - \delta)}{L(p)} \leq D \quad \forall \, p > p_c. \tag{2.1}$$

3. **Theorem [9, Theorem 2]** There are constants $C_0 > 0$ and C_1 such that for all $p > p_c$

$$C_0 P_{cr}\big[0 \leftrightarrow \partial B(L(p))\big] \leq \theta(p) \leq C_1 P_{cr}\big[0 \leftrightarrow \partial B(L(p))\big], \tag{2.2}$$

where $\theta(p) = P_p(0 \leftrightarrow \infty)$ is the percolation function for Bernoulli percolation.

Finally we mention the following result on the behavior of $P_{cr}(0 \leftrightarrow \partial B(n))$. It is believed (see Chapters 9 and 10 in [6] for background) that for $2D$ percolation on sufficiently 'nice' 2D lattices this has a power law (with critical exponent 5/48) but so far this has only been proved for site percolation on the triangular lattice (see [12]). The following is sufficient for our purpose.
There exists a constant D_1 such that

$$\frac{P_{cr}(0 \leftrightarrow \partial B(n))}{P_{cr}(0 \leftrightarrow \partial B(m))} \geq D_1 \sqrt{\frac{m}{n}}, \quad 1 \leq m \leq n. \tag{2.3}$$

For $m = 1$ this was proved in [2, Corollary (3.15)]. For general m it can be proved in a similar way, using a block argument.

2.2   Proof of Theorem 1

Proof. As it is pointed out in [3], it is very easy to see that

$$P(\hat{R} \geq n) \geq P_{cr}(0 \leftrightarrow \partial B(n)), \tag{2.4}$$

since the whole p_c-open cluster of the origin is invaded before *any* edge with τ value larger than p_c is added to the invasion cluster. To prove that the l.h.s. of (2.4) is smaller than some

constant c times the r.h.s. is more involved. First note that it suffices to prove this for the case that n is of the form 2^k. Indeed, if it holds for those special cases then, for any $2^{k-1} < n < 2^k$ we have

$$P(\hat{R} \geq n) \leq P(\hat{R} \geq 2^{k-1}) \leq cP_{cr}\left[0 \longleftrightarrow \partial B(2^{k-1})\right]$$

$$\overset{(2.3)}{\leq} \bar{c}P_{cr}\left[0 \longleftrightarrow \partial B(2^k)\right] \leq \bar{c}P_{cr}\left[0 \longleftrightarrow \partial B(n)\right].$$

First some additional notation and definitions. As in [7] we define $\log^{(0)} k = k$ and $\log^{(j)} k = \log(\log^{(j-1)} k)$ for all $j \geq 1$, as long as the right-hand side is well defined. For $k > 10$ let

$$\log^* k = \min\{j > 0 \ : \ \log^{(j)} k \text{ is well-defined and } \log^{(j)} k \leq 10\}, \tag{2.5}$$

where the choice of the constant 10 is quite arbitrary. Clearly, $\log^{(j)} k > 2$, $j = 0, 1, \ldots, \log^* k$ and $k > 10$. Further,

$$p_k(j) := \inf\left\{p > p_c \ : \ L(p) \leq \frac{2^k}{C_2 \log^{(j)} k}\right\}, \tag{2.6}$$

where the constant C_2 will be chosen later. It is easy to see that $p_k(j)$ is well-defined for all sufficiently large k (in fact, for all k with $2^k > C_2 k$), and that the sequence $\{p_k(j)\}_{j=0}^{\log^* k}$ is decreasing in j. The definition of $p_k(j)$ together with the right continuity of $L(p)$ and (2.1) readily implies that

$$C_2 \log^{(j)} k \leq \frac{2^k}{L(p_k(j))} \leq DC_2 \log^{(j)} k. \tag{2.7}$$

Now we decompose $P(\hat{R} \geq n)$ according to the value of $\hat{\tau}$ as follows, where we note that since τ has a continuous distribution, $\hat{\tau}$ does not coincide with $p_k(j)$ for any $j = 0, \ldots, \log^* k$, almost surely.

$$P(\hat{R} \geq n) = P(\hat{R} \geq n, p_k(0) < \hat{\tau}) + P(\hat{R} \geq n, \hat{\tau} < p_k(\log^* k))$$

$$+ \sum_{j=0}^{\log^* k - 1} P(\hat{R} \geq n, p_k(j+1) < \hat{\tau} < p_k(j)). \tag{2.8}$$

To bound the terms in (2.8), we will use the following observations made in [3]. Let p be an arbitrary number between p_c and 1.

Observations

(a) $\hat{\tau} < p$ if and only if the origin belongs to an infinite p–open cluster.

(b) If $\hat{\tau} > p$ and $\hat{R} \geq n$, then there is a p–closed circuit around O in the dual lattice with diameter at least n.

The event in observation (b) will be denoted by $A_{n,p}$.

$$A_{n,p} := \{\exists \ p\text{-closed circuit around } O \text{ in the dual with diameter at least } n\}.$$

Starting with the first term of (2.8), Observation (b) gives

$$P(\hat{R} \geq n, p_k(0) < \hat{\tau}) \leq P(A_{n,p_k(0)}). \tag{2.9}$$

It is well-known (see [3] for more explanation and references) that there exist C_3 and C_4 such that for all $p > p_c$,

$$P(A_{n,p}) \leq C_3 \exp\left\{ -\frac{C_4 n}{L(p)} \right\} \tag{2.10}$$

Using the lower bound in (2.7) and the definition of $\log^{(0)} k$ we get that

$$P(A_{n,p_k(0)}) \leq C_3 \exp\left\{ -\frac{C_4 n}{L(p_k(0))} \right\} \overset{(2.7)}{\leq} C_3 n^{-C_4 C_2} \tag{2.11}$$

As mentioned above, we have $P_{cr}(0 \leftrightarrow \partial B(n)) \geq Cn^{-1/2}$. Hence, by taking $C_2 \geq 1/C_4$, we can ensure that

$$P(A_{n,p_k(0)}) \leq C_3 n^{-1} \leq \tilde{C}_3 P_{cr}(0 \leftrightarrow \partial B(n)).$$

Remark: *For future purpose we will even take $C_2 \geq 2/C_4$.*

For the second term of (2.8) we apply observation (a) to get

$$P(\hat{R} \geq n, \hat{\tau} < p_k(\log^* k)) \leq P(\hat{\tau} < p_k(\log^* k)) \overset{\text{Obs. (a)}}{\leq} \theta(p_k(\log^* k)).$$

Furthermore, using (2.2), (2.7), the definition of $p_k(\log^* k)$ and (2.3), we have

$$\begin{aligned}
\theta(p_k(\log^* k)) &\leq C_1 P_{cr}\left[0 \leftrightarrow \partial B(L(p_k(\log^* k))\right] \\
&\leq C_1 P_{cr}\left[0 \leftrightarrow \partial B(\frac{2^k}{10DC_2})\right] \\
&\leq C_5 P_{cr}\left[0 \leftrightarrow \partial B(n)\right],
\end{aligned}$$

for some constant C_5.

Now let us consider a typical term in the summation in (2.8). The two observations a few lines below (2.8) (and the definition of $A_{n,p}$) give

$$\begin{aligned}
P(\hat{R} &\geq n, p_k(j+1) < \hat{\tau} < p_k(j)) \\
&\leq P(0 \overset{p_k(j)}{\longleftrightarrow} \infty, A_{n,p_k(j+1)}) \\
&\leq \theta(p_k(j))P(A_{n,p_k(j+1)}),
\end{aligned} \tag{2.12}$$

where in the last inequality we use the Harris-FKG inequality [6, Section 2.2]. To bound the first factor in the right hand side of (2.12), note that

$$\begin{aligned}
\theta(p_k(j)) &\overset{(2.2)}{\leq} C_1 P_{cr}(0 \leftrightarrow L(p_k(j))) \\
&= C_1 P_{cr}(0 \leftrightarrow \partial B(2^k)) \frac{P_{cr}(0 \leftrightarrow L(p_k(j)))}{P_{cr}(0 \leftrightarrow \partial B(2^k))} \\
&\overset{(2.3)}{\leq} \frac{C_1}{D_1} P_{cr}(0 \leftrightarrow \partial B(2^k)) \left(\frac{2^k}{L(p_k(j))}\right)^{1/2} \\
&\overset{(2.7)}{\leq} \frac{C_1}{D_1} P_{cr}(0 \leftrightarrow \partial B(2^k))(DC_2 \log^{(j)} k)^{1/2}.
\end{aligned} \tag{2.13}$$

The second factor in the right hand side of (2.12) can be bounded using (2.10), (2.7), (2.6) and the choice of C_2:

$$P(A_{n,p_k(j+1)}) \leq C_3 \exp\left\{-\frac{C_4 n}{L(p_k(j+1))}\right\} \leq C_3(\log^{(j)} k)^{-1}, \tag{2.14}$$

Combining (2.13) and (2.14) gives

$$\theta(p_k(j))P(A_{n,p_k(j+1)}) \leq C_8(\log^{(j)} k)^{-1/2} P_{cr}\left[0 \leftrightarrow \partial B(n)\right]. \tag{2.15}$$

To conclude the proof it suffices to show that

$$\sup_{k>10} \sum_{j=0}^{\log^* k - 1} (\log^{(j)} k)^{-1/2} < \infty. \tag{2.16}$$

Recall from the definitions that $\log^{(j)} k > 2$. Applying this to the case $j = \log^* k$ shows that the last term in the sum in (2.16) is at most $(e^2)^{-1/2}$. Similarly, the penultimate term is at most $(\exp(e^2))^{-1/2}$, etc. This leads to the finite upper bound $C_9 := \frac{1}{\sqrt{e^2}} + \frac{1}{\sqrt{e^{e^2}}} + \ldots$ for the l.h.s. of (2.16).

Putting everything together we get

$$P(\hat{R} \geq n) \leq (\tilde{C}_3 + C_5 + C_8 C_9) P_{cr}\left[0 \leftrightarrow \partial B(n)\right].$$

$\square$

2.3   Proof of Theorem 2

For short, we use the following notation:

$$\pi(n) = P_{cr}(0 \leftrightarrow \partial B(n));$$
$$\pi(n,p) = P_p(0 \leftrightarrow \partial B(n)).$$

Recall that $s(n) = n^2 \pi(n)$.

The difficult part of Theorem 2 is the third inequality. We need the following key ingredient.

Lemma 4. *There exist constants C_{10} and C_{11}, such that*

$$P_p\left(0 \leftrightarrow \infty, |\mathcal{C}(0) \cap B(2^k)| > s(n)\right)$$
$$\leq \theta(p) \, 2C_{10} \exp\left\{-(2C_{11})^{-1}\frac{s(n)}{2^{2k}\pi(2^k,p)}\right\}, \quad p > p_c, 2^k \leq n.$$

Proof. The proof is based on the following moment estimate:

$$E_p\left(|\mathcal{C}(0) \cap B(2^k)|^t \,\Big|\, 0 \longleftrightarrow \infty\right) \leq C_{10} \, t! \left[C_{11} \, 2^{2k} \, \pi(2^k,p)\right]^t, \quad t \geq 1. \tag{2.17}$$

Very similar estimates were proved in [8, Theorem (8)] and in [11]. To adapt their proofs in order to obtain (2.17), one merely needs that the inequality $\sum_{m=0}^{n} \pi(n,p) \leq Cn\pi(n,p)$ holds for all $p \geq p_c$ (with some constant C independent of p). From (2.17), we readily get

$$E_p\left(\exp\left\{\lambda\frac{|\mathcal{C}(0) \cap B(2^k)|}{2^{2k}\,\pi(2^k,p)}\right\} \,\Big|\, 0 \longleftrightarrow \infty\right) \leq C_{10}\frac{1}{1 - \lambda C_{11}}, \quad 0 < \lambda < C_{11}^{-1}.$$

Taking $\lambda = (2C_{11})^{-1}$ we easily obtain the estimate of the lemma. $\square$

Proof of Theorem 2. The first inequality follows from [8, Remark (9)]. The second inequality follows immediately from the fact that the p_c-open cluster containing the origin is a subset of $\hat{V}$.

The third inequality will be proved by a decomposition, somewhat similar to the one in Theorem 1, but now two-fold: this time we will also decompose according to the value of $\hat{R}$. As in the proof of Theorem 1, without loss of generality we may assume that n is of the form 2^N.

Let

$$E_{n,k} = \{2^{k-1} < \hat{R} \le 2^k, |\hat{V}| > s(n)\}.$$

Note that $s(n) \ge C_{12} n^{3/2}$, and $|B(2^k)| \le C_{13} 2^{2k}$. Letting

$$k_0 := \max\{k : C_{13} 2^{2k} \le C_{12} n^{3/2}\},$$

for $k < k_0$, $\hat{R} \le 2^k$ implies $|\hat{V}| \le C_{13} 2^{2k} \le s(n)$, and hence $E_{n,k} = \emptyset$. Therefore, we can write

$$P(|\hat{V}| > s(n)) \le P(\hat{R} > n) + \sum_{k=k_0}^{N} P(E_{n,k}). \tag{2.18}$$

The first term on the right hand side is at most $C_{14}\pi(n)$, by Theorem 1. Consider now a general term of the sum. We decompose this according to the value of $\hat{\tau}$ as follows:

$$
\begin{aligned}
P(2^{k-1} &< \hat{R} \le 2^k, |\hat{V}| > s(n)) \\
&= P(E_{n,k}, \hat{\tau} > p_k(0)) + \sum_{j=0}^{\log^* k} P(E_{n,k}, p_k(j+1) < \hat{\tau} < p_k(j)),
\end{aligned}
\tag{2.19}
$$

where we let $p_k(\log^* k + 1) = p_c$.

We first look at the event in the first term on the right hand side. This event implies the occurrence of $A_{2^{k-1}, p_k(0)}$. Hence, by virtue of (2.11), its probability is at most $C_{15}(2^{2k})^{-\tilde{C}_4 C_2}$. By the choice of C_2, we have $\tilde{C}_4 C_2 \ge 1$. Hence the sum over $k_0 \le k \le N$ is bounded by $C_{16}(2^{2k_0})^{-1}$. By the definition of k_0, this is $o(\pi(n))$.

Consider now the event in the general term on the right hand side of (2.19). This event implies the following two events:

(i) $A_{2^{k-1}, p_k(j+1)}$;

(ii) $\{0 \overset{p_k(j)}{\longleftrightarrow} \infty, |\mathcal{C}(0; p_k(j)) \cap B(2^k)| > s(n)\}$;

where $\mathcal{C}(0; p)$ denotes the p-open cluster of 0. Since (i) is a decreasing and (ii) an increasing event, the Harris-FKG inequality yields that the general term in (2.19) is at most the product of the probabilities of event (i) and event (ii).

As to event (i), the same arguments that led to (2.14) (and noting the Remark a few lines below (2.11)) show that for $j < \log^* k$ this has probability less than or equal to

$$C_3(\log^{(j)} k)^{-1} \tag{2.20}$$

It is easy to see that, after increasing the value of C_3 if necessary, this bound even holds for $j = \log^* k$.

As to event (ii), by Lemma 4 this has probability at most

$$\theta(p_k(j))(2C_{10}) \exp\left\{-(2C_{11})^{-1}\frac{s(n)}{2^{2k}\pi(2^k, p_k(j))}\right\}. \tag{2.21}$$

Applying the first inequality in (2.2) to the probability in the exponent in (2.21), and then applying (2.13) twice, shows that (2.21) is at most a constant times

$$\pi(2^k)(\log^{(j)} k)^{1/2} \exp\left\{-C_{18}\frac{2^{2N}\pi(n)}{2^{2k}\pi(2^k)}(\log^{(j)} k)^{-1/2}\right\}. \tag{2.22}$$

Combining this with (2.20) gives that the general term in (2.19) is at most

$$C_{17}\pi(n)(\log^{(j)} k)^{-1/2}\frac{\pi(2^k)}{\pi(n)} \exp\left\{-C_{18}\frac{2^{2N}\pi(n)}{2^{2k}\pi(2^k)}(\log^{(j)} k)^{-1/2}\right\}. \tag{2.23}$$

Due to (2.3), this as at most

$$C_{19}\pi(n)(\log^{(j)} k)^{-1/2}2^{(N-k)/2} \exp\left\{-C_{20}2^{(N-k)(3/2)}(\log^{(j)} k)^{-1/2}\right\}. \tag{2.24}$$

We split the sums over j and k into two parts:

(1) $2^{(N-k)} \leq (\log^{(j)} k)^{1/2}$;

(2) $2^{(N-k)} > (\log^{(j)} k)^{1/2}$.

In case (1), we bound the exponential in (2.24) by 1, and we have

$$(\log^{(j)} k)^{-1/2}2^{(N-k)/2} \leq (\log^{(j)} k)^{-1/4} \leq C_{21}(\log^{(j)} N)^{-1/4}.$$

The number of possible values of k is at most

$$(2\log 2)^{-1}\log^{(j+1)} k \leq C_{22}(\log^{(j)} N)^{1/8}.$$

Hence the contribution of this case is bounded by

$$\sum_{j=0}^{\log^* N}(\log^{(j)} N)^{-1/8} \leq C_{23}.$$

In case (2), we bound the exponential by $\exp\{-C_{20}2^{(N-k)/2}\}$, and we have

$$(\log^{(j)} k)^{-1/2}2^{(N-k)/2} \leq 2^{(N-k)/2}.$$

The sum over k can be bounded as follows:

$$\sum_{k:N-k\geq c\log^{(j+1)} N}2^{(N-k)/2} \exp\{-C_{20}2^{(N-k)/2}\} \leq C_{24}\exp\{-C_{25}(\log^{(j)} N)^{c_1}\},$$

for some $c_1 > 0$. The sum of the right hand side over j is again bounded. This proves the theorem. $\qquad\square$

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