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## A CORRELATION INEQUALITY FOR CONNECTION EVENTS IN PERCOLATION

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It is well-known in percolation theory (and intuitively plausible) that two events of the form "there is an open path from s to a" are positively correlated. We prove the (not intuitively obvious) fact that this is still true if we condition on an event of the form "there is no open path from s to t."

**1. Introduction and statement of results.** We consider the usual bond percolation models on a (finite or countably infinite) graph G = (V, E): each  $e \in E$  is "open" (has value 1) with probability p(e) and "closed" (has value 0) with probability 1 - p(e), independently of all other edges. We write P for the corresponding probability distribution on  $\Omega := \{0, 1\}^E$ . For general background see [4].

For  $s, a \in V$  we write  $s \leftrightarrow a$  for the event that there is an open path from s to a, and  $s \nleftrightarrow a$  for the complementary event.

Positive (i.e., nonnegative) correlation of any two events  $s \leftrightarrow a$  and  $s \leftrightarrow b$  follows from Harris' inequality [5] (Theorem 2.1 below). The correlation inequality of the title says that this phenomenon persists if we condition on any event  $s \leftrightarrow t$ .

THEOREM 1.1. For any  $s, a, b, t \in V$ ,  $P(s \leftrightarrow a, s \leftrightarrow b|s \leftrightarrow t) \ge P(s \leftrightarrow a|s \leftrightarrow t)P(s \leftrightarrow b|s \leftrightarrow t).$ 

The intuition for this is not very clear. In particular it is *not* true if we condition on  $s \leftrightarrow t$  rather than  $s \leftrightarrow t$ . (Consider the graph with vertices s, a, b, t and each of s, t joined to each of a, b.)

From now on we fix  $s \in V$ , and set, for  $X \subseteq V$ ,  $Q_X = \{s \leftrightarrow x \forall x \in X\}$  and  $R_X = \{s \leftrightarrow x \forall x \in X\}$ .

THEOREM 1.2. For any  $A, B, X, Y \subseteq V$ ,

(1)  $P(Q_A R_X) P(Q_B R_Y) \le P(Q_{A \cup B} R_{X \cap Y}) P(R_{X \cup Y}).$ 

**REMARKS** 1. Of course we recover Theorem 1.1 from Theorem 1.2 by taking  $A = \{a\}, B = \{b\}$  and  $X = Y = \{t\}$ . This is not generalization for its own sake: the more general form is needed for the proof.

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- 2. The perhaps intuitively more natural statement obtained by replacing  $R_{X\cup Y}$  by  $Q_{A\cap B}R_{X\cup Y}$  in Theorem 1.2 is *not* true: take  $V(G) = \{s, x, y, a\}$ ,  $E(G) = \{sx, xa, ay, ys\}$  and  $X = \{x\}$ ,  $Y = \{y\}$ ,  $A = B = \{a\}$ .
- 3. As pointed out to us by the referee, Theorem 1.1 can be generalized to sets of vertices S, A, B, T by replacing s by  $S, \ldots, t$  by T, and interpreting  $X \leftrightarrow Y$  as  $\{\exists x \in X, y \in Yx \leftrightarrow y\}$ . To see this, simply identify all vertices in each of S, A, B, T, retaining multiple edges, and apply Theorem 1.1.
- 4. Note that if we replace A by  $A \setminus B$  in Theorem 1.2, the r.h.s. of (1) remains the same and the l.h.s. does not decrease. So Theorem 1.2 as stated above is not more general than the case  $A \cap B = \emptyset$ .
- 5. The original motivation for Theorem 1.1 was a conjecture we learned from the late P. W. Kasteleyn (personal communication, circa 1985), a slightly informal description of which is as follows. Let G = (V, E) be a finite graph, W some subset of V and G̃ = (Ṽ, Ẽ) a copy of G. For each e ∈ E and v ∈ V, let ẽ and ṽ be the corresponding edge and vertex in G̃, respectively. Now we "glue" G and G̃ together by identifying w with w̃ for w ∈ W, and on this new graph consider any percolation model with p(ẽ) = p(e) for all e ∈ E. The conjecture is then that, for every a, b ∈ V, P(a ↔ b) ≥ P(a ↔ b̃). There is in fact a slight concrete connection with Theorem 1.1, in that a special case of the latter says that when |W| = 2, say W = {v, w}, one has P(a ↔ b|v ↔ w) ≥ P(a ↔ b̃|v ↔ w). But we feel that Theorem 1.1 is more interesting for its own sake and believe it has potential applications in percolation theory in general.

**2. Background.** We just recall the two correlation inequalities we will need in Section 3. For more extensive discussions see [2].

An event  $\mathscr{A}$  (i.e., a subset of  $\Omega$ ) is called *increasing* if  $\mathscr{A} \ni \omega \leq \omega'$  implies  $\omega' \in A$ . (Here  $\omega \leq \omega'$  means  $\omega_e \leq \omega'_e$  for all  $e \in E$ .) The following correlation inequality is due to Harris [5].

THEOREM 2.1. For any increasing  $\mathscr{A}, \mathscr{B} \subset \Omega$ ,

 $P(\mathscr{A}\mathscr{B}) \geq P(\mathscr{A})P(\mathscr{B}).$ 

Of course this is equivalent to saying that for any increasing  $\mathscr{A}$  and *decreasing*  $\mathscr{B}$ ,  $P(\mathscr{AB}) \leq P(\mathscr{A})P(\mathscr{B})$ .

There are a number of significant extensions of Harris' inequality, notably that of Fortuin, Kasteleyn and Ginibre [3]. (We are informed by the referee that the inequality was essentially given a bit earlier in [6].) Our main tool is the considerably more general Ahlswede–Daykin (or "four functions") theorem [1], namely:

THEOREM 2.2. Let N be a finite set and let  $\mathscr{P}(N)$  denote the set of all subsets of N. Suppose  $\alpha, \beta, \gamma, \delta$ :  $\mathscr{P}(N) \to \mathbf{R}^+$  satisfy

(2) 
$$\alpha(S)\beta(T) \le \gamma(S \cap T)\delta(S \cup T) \quad \forall S, T \le N.$$

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Then  $\sum \alpha(S) \sum \beta(S) \leq \sum \gamma(S) \sum \delta(S)$  (where the sums are over all  $S \subseteq N$ ).

**3. Proof of Theorem 1.2.** We assume *G* is finite. (If *G* is countably infinite, the result follows from the finite case by obvious limit arguments.) The proof is by induction on the number of vertices |V|. If |V| = 1, the result is trivial. Suppose it always holds if  $|V| \le n$  and consider a graph *G* with n + 1 vertices.

Set  $X \cap Y = Z$ . If  $Z = \emptyset$  then (1) follows from the Harris inequality:

$$\begin{split} P(Q_A R_X) P(Q_B R_Y) &\leq P(Q_A) P(R_X) P(Q_B) P(R_Y) \\ &\leq P(Q_A Q_B) P(R_X R_Y) \\ &= P(Q_{A \cup B} R_{X \cap Y}) P(R_{X \cup Y}). \end{split}$$

If  $Z \neq \emptyset$  we proceed as follows: set  $N = \{y \notin Z: y \sim Z\}$  (where  $y \sim Z$  means y is adjacent to at least one vertex of Z). Define the (random) set

 $\mathbf{S} = \{ y \in N : \text{ there is an open edge from } y \text{ to } Z \}.$ 

We use *S*, *T* for possible values of **S** and write P(S) for P(S = S) and  $P(\cdot|S)$  for the conditional distribution given S = S. We may expand

$$P(Q_A R_X) = \sum_S P(S) P(Q_A R_X | S)$$

(where the sum is over all subsets of N) and similarly for the other terms in (1). Thus if we define

$$\begin{split} &\alpha(S) = P(S)P(Q_A R_X|S),\\ &\beta(S) = P(S)P(Q_B R_Y|S),\\ &\gamma(S) = P(S)P(Q_{A\cup B} R_{X\cap Y}|S),\\ &\delta(S) = P(S)P(R_{X\cup Y}|S), \end{split}$$

then (1) becomes

$$\sum lpha(S) \sum eta(S) \leq \sum \gamma(S) \sum \delta(S),$$

where *S* runs over the subsets of *N*. Theorem 2.2 says that to verify this we just need to establish (2), which, since (as one can easily check)  $P(S)P(T) = P(S \cup T)P(S \cap T)$ , is the same as

$$(3) \qquad P(Q_A R_X | S) P(Q_B R_Y | T) \leq P(Q_{A \cup B} R_{X \cap Y} | S \cap T) P(R_{X \cup Y} | S \cup T).$$

Let P' refer to the percolation model for the graph G', obtained from G by removing Z, with edge probabilities as in our original percolation model on G. Then it is easy to see that for any  $C, W \subseteq V \setminus Z$  and  $S \subseteq N$ ,

(4) 
$$P(Q_C R_{W \cup Z} | S) = P'(Q_C R_{W \cup S}).$$

Now we obtain (3) as follows: Let  $X' = X \setminus Z$  and  $Y' = Y \setminus Z$ . We have

$$\begin{split} P(Q_A R_X | S) P(Q_B R_Y | T) &= P'(Q_A R_{X' \cup S}) P'(Q_B R_{Y' \cup T}) \\ &\leq P'(Q_{A \cup B} R_{(X' \cup S) \cap (Y' \cup T)}) P'(R_{(X' \cup S) \cup (Y' \cup T)}) \\ &\leq P'(Q_{A \cup B} R_{(S \cap T)}) P'(R_{(X' \cup Y') \cup (S \cup T)}) \\ &= P(Q_{A \cup B} R_{X \cap Y} | S \cap T) P(R_{X \cup Y} | S \cup T), \end{split}$$

where the first equality follows from applying (4) twice (with W = X' and W = Y', respectively), the first inequality from the induction hypothesis [which says that (1) holds for G'], the second inequality from  $(S \cap T) \subseteq (X' \cup S) \cap (Y' \cup T)$ , and the second equality from again applying (4) twice (with  $W = \emptyset$  and  $W = X' \cup Y'$ , respectively).  $\Box$ 

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