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# A Correlation Inequality for Connection Events in Percolation 

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#### Abstract

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 $t$ ".


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## 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 [3].

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

Positive (i.e. nonnegative) correlation of any two events $s \longleftrightarrow a$ and $s \longleftrightarrow b$ 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 $s \longleftrightarrow t$.

Theorem 1.1 For any $s, a, b, t \in V$

$$
P(s \longleftrightarrow a, s \longleftrightarrow b \mid s \longleftrightarrow t) \geq P(s \longleftrightarrow a \mid s \longleftrightarrow t) P(s \longleftrightarrow b \mid s \longleftrightarrow t) .
$$

The intuition for this is not very clear. In particular it is not true if we condition on $s \longleftrightarrow t$ rather than $s \longleftrightarrow t$. (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 $X \subseteq V, Q_{X}=\{s \longleftrightarrow x \forall x \in X\}$ and $R_{X}=\{s \nleftarrow x \forall x \in X\}$.

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

$$
\begin{equation*}
P\left(Q_{A} R_{X}\right) P\left(Q_{B} R_{Y}\right) \leq P\left(Q_{A \cup B} R_{X \cap Y}\right) P\left(R_{X \cup Y}\right) \tag{1}
\end{equation*}
$$

## 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.
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 $V(G)=\{s, x, y, a\}, E(G)=\{s x, x a, a y, y s\}$ and $X=\{x\}, Y=\{y\}, A=B=\{a\}$.
3. Note that if we replace $A$ by $A \backslash B$ 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 $A \cap B=\emptyset$.
4. 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 $\tilde{G}=(\tilde{V}, \tilde{E})$ a copy of $G$. For each $e \in E$ and $v \in V$, let $\tilde{e}$ and $\tilde{v}$ be the corresponding edge and vertex in $\tilde{G}$ respectively. Now we 'glue' $G$ and $\tilde{G}$ together by identifying $w$ with $\tilde{w}$ for $w \in W$, and on this new graph consider any percolation model with $p(\tilde{e})=p(e)$ for all $e \in E$. The conjecture is then that, for every $a, b \in V, P(a \longleftrightarrow b) \geq P(a \longleftrightarrow \tilde{b})$. 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 $P(a \longleftrightarrow b \mid v \longleftarrow w) \geq P(a \longleftrightarrow \tilde{b} \mid v \longleftrightarrow w)$. 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 $\mathcal{A}$ (i.e. a subset of $\Omega$ ) is called increasing if $\mathcal{A} \ni \omega \leq \omega^{\prime}$ implies $\omega^{\prime} \in A$. (Here $\omega \leq \omega^{\prime}$ means $\omega_{e} \leq \omega_{e}^{\prime}$ for all $e \in E$ ). The following correlation inequality is due to Harris [5].

Theorem 2.1 For any increasing $\mathcal{A}, \mathcal{B} \subset \Omega$,

$$
P(\mathcal{A B}) \geq P(\mathcal{A}) P(\mathcal{B})
$$

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

There are a number of significant extensions of Harris' inequality, notably that of Fortuin, Kasteleyn and Ginibre [4]. Our main tool is the considerably more general Ahlswede-Daykin (or "Four Functions") Theorem [1], viz.

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

$$
\begin{equation*}
\alpha(S) \beta(T) \leq \gamma(S \cap T) \delta(S \cup T) \quad \forall S, T \subseteq N \tag{2}
\end{equation*}
$$

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 $|V| \leq n$ and consider a graph $G$ with $n+1$ vertices.

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

$$
\begin{aligned}
P\left(Q_{A} R_{X}\right) P\left(Q_{B} R_{Y}\right) & \leq P\left(Q_{A}\right) P\left(R_{X}\right) P\left(Q_{B}\right) P\left(R_{Y}\right) \\
& \leq P\left(Q_{A} Q_{B}\right) P\left(R_{X} R_{Y}\right) \\
& =P\left(Q_{A \cup B} R_{X \cap Y}\right) P\left(R_{X \cup Y}\right) .
\end{aligned}
$$

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 $\mathbf{S}$ and write $P(S)$ for $P(\mathbf{S}=S)$ and $P(\cdot \mid S)$ for the conditional distribution given $\mathbf{S}=S$. We may expand

$$
P\left(Q_{A} R_{X}\right)=\sum_{S} P(S) P\left(Q_{A} R_{X} \mid S\right)
$$

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

$$
\begin{aligned}
\alpha(S) & =P(S) P\left(Q_{A} R_{X} \mid S\right) \\
\beta(S) & =P(S) P\left(Q_{B} R_{Y} \mid S\right) \\
\gamma(S) & =P(S) P\left(Q_{A \cup B} R_{X \cap Y} \mid S\right) \\
\delta(S) & =P(S) P\left(R_{X \cup Y} \mid S\right),
\end{aligned}
$$

then (1) becomes

$$
\sum \alpha(S) \sum \beta(S) \leq \sum \gamma(S) \sum \delta(S)
$$

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

$$
\begin{equation*}
P\left(Q_{A} R_{X} \mid S\right) P\left(Q_{B} R_{Y} \mid T\right) \leq P\left(Q_{A \cup B} R_{X \cap Y} \mid S \cap T\right) P\left(R_{X \cup Y} \mid S \cup T\right) \tag{3}
\end{equation*}
$$

Let $P^{\prime}$ refer to the percolation model for the graph $G^{\prime}$, 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 $C, W \subseteq V \backslash Z$ and $S \subseteq N$,

$$
\begin{equation*}
P\left(Q_{C} R_{W \cup Z} \mid S\right)=P^{\prime}\left(Q_{C} R_{W \cup S}\right) \tag{4}
\end{equation*}
$$

Now we obtain (3) as follows: Let $X^{\prime}=X \backslash Z$ and $Y^{\prime}=Y \backslash Z$. We have

$$
\begin{aligned}
P\left(Q_{A} R_{X} \mid S\right) P\left(Q_{B} R_{Y} \mid T\right) & =P^{\prime}\left(Q_{A} R_{X^{\prime} \cup S}\right) P^{\prime}\left(Q_{B} R_{Y^{\prime} \cup T}\right) \\
& \leq P^{\prime}\left(Q_{A \cup B} R_{\left(X^{\prime} \cup S\right) \cap\left(Y^{\prime} \cup T\right)}\right) P^{\prime}\left(R_{\left(X^{\prime} \cup S\right) \cup\left(Y^{\prime} \cup T\right)}\right) \\
& \leq P^{\prime}\left(Q_{A \cup B} R_{(S \cap T)}\right) P^{\prime}\left(R_{\left(X^{\prime} \cup Y^{\prime}\right) \cup(S \cup T)}\right) \\
& =P\left(Q_{A \cup B} R_{X \cap Y} \mid S \cap T\right) P\left(R_{X \cup Y} \mid S \cup T\right)
\end{aligned}
$$

where the first equality follows from applying (4) twice (with $W=X^{\prime}$ and $W=Y^{\prime}$ respectively), the first inequality from the induction hypothesis (which says that (1) holds for $G^{\prime}$ ), the second inequality from $(S \cap T) \subseteq\left(X^{\prime} \cup S\right) \cap\left(Y^{\prime} \cup T\right)$, and the second equality from again applying (4) twice (with $W=\emptyset$ and $W=X^{\prime} \cup Y^{\prime}$ respectively).

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