Some percolations involving the Gaussian free fields
Nathalie Eisenbaum

To cite this version:
Nathalie Eisenbaum. Some percolations involving the Gaussian free fields. Electronic Communications in Probability, 2021, 26, pp.1-8. 10.1214/21-ECP379. hal-03400953

HAL Id: hal-03400953
https://hal.science/hal-03400953
Submitted on 25 Oct 2021

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers.

L’archive ouverte pluridisciplinaire HAL, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d’enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.
Some percolations involving the Gaussian free fields

Nathalie Eisenbaum

Abstract: Consider an infinite, connected, locally finite graph with vertex set \( V \). Intuitively a simple point process on \( V \) with attractive properties, should percolate more easily than a Bernoulli point process with the same marginals. Although it seems wrong to imagine that it could be true in general, we confirm this intuition on several examples involving Gaussian free fields and permanental free fields.

Keywords: Percolation, Gaussian free field, permanental process, positive association, Markov chain, local times.

MSC2010 subject classifications: 60A10, 60G10, 60G15, 60G17, 60G50, 60G55, 60G60

1 Introduction

Consider a non-oriented, infinite, connected, locally finite, graph \( G \), with vertex set \( V \) and edge set \( E \). Given a family of Bernoulli variables \( (Y_x, x \in V) \), one may ask whether the random subgraph of \( G \) with vertex set \( \{x \in V : Y_x = 1\} \) and edge set \( \{[x, y] \in E : Y_x = 1 \text{ and } Y_y = 1\} \), contains an infinite connected component. In short does \( \{x \in V : Y_x = 1\} \) percolate?

There is a general answer to this question in the case when the variables \( Y_x, x \in V \) are i.i.d. One sets: \( p = \mathbb{P}[Y_x = 1] \). There exists a critical probability \( p_c^{\text{site}}(G) \) in \( [0, 1] \), such that for \( p > p_c^{\text{site}}(G) \), \( \{x \in V : Y_x = 1\} \) percolates and for \( p < p_c^{\text{site}}(G) \), \( \{x \in V : Y_x = 1\} \) does not.

For \( G = \mathbb{Z}^d \), one knows that \( p_c^{\text{site}}(\mathbb{Z}^2) > 1/2 \) (see [14]) and that for \( d \geq 3: p_c^{\text{site}}(\mathbb{Z}^d) < 1/2 \) (see [7]).

When the Bernoulli variables are not independent, it is much more difficult to solve this question. As an example of dependent Bernoulli variables, take: \( Y_x = 1_{\{\eta_x > h\}} \), \( x \in \mathbb{Z}^d \), where \( h \) is a fixed real level and \( (\eta_x, x \in \mathbb{Z}^d) \) is a centered Gaussian process. One particular case has been intensively studied: the case when \( \eta \) is the Gaussian free
field on $\mathbb{Z}^d$ for $d \geq 3$, which means that $\eta$ is a centered Gaussian process with the following covariance:

$$\mathbb{E}[\eta_x \eta_y] = \mathbb{E}_x \sum_{n=0}^{\infty} 1\{U_n=y\}$$

where $(U_n)_{n \geq 0}$ is a simple random walk on $\mathbb{Z}^d$. By definition, this covariance is the Green function of $(U_n)_{n \geq 0}$.

There exists a critical level $h_\ast$ such that

- for $h < h_\ast$, the set $\{x \in \mathbb{Z}^d : \eta_x > h\}$ a.s. has an infinite connected component.
- for $h > h_\ast$, the set $\{x \in \mathbb{Z}^d : \eta_x > h\}$ a.s. has only finite connected component.

It has been shown by Bricmont, Lebowitz and Maes [5], that $h_\ast \geq 0$ and that moreover in dimension 3: $h_\ast < \infty$. Then Rodriguez and Sznitman [20] have shown that in any dimension $h_\ast > 0$ (i.e. $\mathbb{P}(\eta_0 > h_\ast) < 1/2$). In [8], the authors suggest that their result could be the consequence of the following conjecture:

$$\mathbb{P}(\eta_0 > h_\ast) < p_{\text{site}}^{\text{site}}(\mathbb{Z}^d),$$

(1.1)

based on the intuition that positive correlation should help in forming clusters and hence an infinite cluster. Indeed the Gaussian free field $\eta$ is positively correlated in the sense that its covariance is positive. But thanks to Pitt [19], this fact implies the much stronger following property called “positive association”:

$$\mathbb{E}[F(\eta)H(\eta)] \geq \mathbb{E}[F(\eta)]\mathbb{E}[H(\eta)]$$

(1.2)

for any couple $(F, H)$ of increasing functionals on functions from $\mathcal{V}$ into $\mathbb{R}$ (increasing with respect to each coordinate). For example, one has for every $x_1, \ldots, x_n$ in $\mathbb{Z}^d$:

$$\mathbb{P}[\eta_{x_1} > h, \ldots, \eta_{x_n} > h] \geq \mathbb{P}[\eta_{x_1} > h]\ldots\mathbb{P}[\eta_{x_n} > h],$$

which legitimates (1.1).

We test this intuition on another positively associated process: $(|\eta_x|, x \in \mathbb{Z}^d)$. We know that $(|\eta_x|, x \in \mathbb{Z}^d)$ is positively associated because $(\eta^2_x, x \in \mathbb{Z}^d)$ is infinitely divisible (see [11]) and is hence positively associated in the sense of (1.2) thanks to [6].

Rodriguez [21] has proved that there exists $h_c < \infty$ such that for $h > h_c$ a.s. the set $\{x \in \mathbb{Z}^d : |\eta_x| > h\}$ does not percolate. One obviously has: $h_\ast \leq h_c$. Consequently: $0 < h_c < \infty$. In view of the above intuition, one would expect that:

$$\mathbb{P}[|\eta_0| > h_c] < p_{\text{site}}^{\text{site}}(\mathbb{Z}^d).$$

We have the following result:

$$\mathbb{P}[|\eta_0| > h_c \sqrt{\mathbb{E}[\eta_0^2]}] \leq p_{\text{site}}^{\text{site}}(\mathbb{Z}^d),$$

which gives the following lower bound.
Theorem 1.1 For $d \geq 3$, one has
\[ h_o \leq h_c < \infty \]
where for $N_o$ real standard Gaussian variable, $h_o$ satisfies: $\mathbb{P}[|N_o| > h_o] = p_c^{\text{site}}(\mathbb{Z}^d)$.

Since one has: $\mathbb{E}[\eta_0^2] \geq 1$, the above result is weaker than its intuition. But as shown in section 2 (Theorem 2.1), this result is “universal” in the sense that it is available for every centered Gaussian process associated to a symmetric transient homogenous Markov chain on any infinite connected graph. In Theorem 2.1 the jumps of the Markov chain are not limited to the closest neighbors.

One can extend Theorem 2.1 from absolute value of Gaussian free fields to permanental free fields. This extension consists in relaxing the assumption of symmetry for the associated Markov chain. We remind that the permanental free fields are also positively associated [12]. The result is presented in section 2 (Theorem 2.2).

Finally, in section 3, we extend the result of Bricmont, Lebowitz and Maes [5] to transient simple symmetric random walks on any regular graph. This extension has been already noticed by Abächerli and Sznitman (Proposition A2 in [1]). In [9], Drewitz, Prévost and Rodriguez go further by showing that $h_* > 0$ for a large class of graphs. The interest of our proof is located in the use of a basic Dynkin isomorphism type theorem.

2 Extension to all Gaussian and Permanental free fields

For $G$ non-oriented, locally finite, infinite connected graph determined by $(\mathcal{V}, E)$, let $U = (U_n)_{n \geq 0}$ be a transient homogenous Markov chain on $\mathcal{V}$, with transition matrix $P = (P(x, y))_{(x, y) \in \mathcal{V} \times \mathcal{V}}$. The Markov chain $U$ is allowed to have jumps from $x$ to $y$ with $[x, y]$ outside $E$. The paths of $(U_n)_{n \geq 0}$ are not necessarily subgraphs of $G$. Actually $U$ lives on the graph $G(U)$ with edge set $E(U) = \{[x, y] \in \mathcal{V} \times \mathcal{V} : P(x, y) + P(y, x) > 0\}$.

We draw attention on a result of Benjamini and Hermon (Theorem 2 in [4]) according to which if $U$ is irreducible, transient and $E(U) = E$ (i.e. $G(U) = G$) then the simple random walk on $G$ might not be transient. More precisely, if $U$ has the additional property to cover $G$ with positive probability, then the simple random walk on $G$ is recurrent.

Since we are dealing only with homogenous Markov chain, we will omit the term homogenous.
Theorem 2.1 Let $G$ be an infinite connected graph with a locally finite vertex set $V$ and edge set $E$. Let $h_G$ be the nonnegative number such that:

$$\mathbb{P}[|N_o| > h_G] = p_c^{site}(G)$$

where $N_o$ is a standard Gaussian real variable. Then for every $h < h_G$, for every centered Gaussian field $(\eta_x)_{x \in V}$ with covariance the Green function of a transient symmetric Markov chain on $V$, the subgraph of $G$ with edge set $\{[x,y] \in E : |\eta_x| > h \text{ and } |\eta_y| > h\}$ has a.s. an infinite connected component.

In case $G$ is such that $p_c^{site}(G) = 1$, Theorem 2.1 does not bring any information on the percolation properties of the absolute value of the Gaussian free fields.

One can extend Theorem 2.1 from absolute value of Gaussian free fields to permanental free fields. This extension consists in relaxing the assumption of symmetry for the associated Markov chain.

We first recall that a permanental process $(\varphi(x), x \in V)$ with index $\beta > 0$ and a kernel $k = (k(x,y), (x,y) \in V \times V)$ is a nonnegative process with finite dimensional Laplace transforms satisfying, for every $x_1, x_2, \ldots, x_n$ in $V$:

$$\mathbb{E}[\exp\left(-\frac{1}{2} \sum_{i=1}^{n} \alpha_i \varphi(x_i)\right)] = \det(I + \alpha K)^{-\beta}$$

where $\alpha$ is the diagonal matrix with diagonal entries $(\alpha_i)_{1 \leq i \leq n}$, $I$ is the $n \times n$-identity matrix and $K$ is the matrix $(k(x_i, x_j))_{1 \leq i,j \leq n}$.

Note that the kernel of a permanental process is not unique.

In case $\beta = 1/2$ and $k$ is symmetric positive semi-definite, $(\varphi_x, x \in V)$ equals in law $(\eta^2_x, x \in V)$ where $(\eta_x, x \in V)$ is a centered Gaussian process with covariance $k$.

Consider a Markov chain with state space $V$ and finite Green function $(g(x,y), (x,y) \in V \times V)$. For every $\beta > 0$, there exists a permanental process with index $\beta$ and kernel $(g(x,y), (x,y) \in V \times V)$ (see [12]). The permanental processes obtained that way, can be called, by analogy with the Gaussian free fields, permanental free fields. Note that the permanental free fields are infinitely divisible and hence positively associated (by using again [6]).

Theorem 2.2 Let $G$ be an infinite connected graph with a locally finite vertex set $V$ and edge set $E$. Let $h_G$ be the nonnegative number such that:

$$\mathbb{P}[|N_o| > h_G] = p_c^{site}(G)$$

where $N_o$ is a standard Gaussian real variable. Then for every $h < h^2_G$, for every permanental field $(\varphi_x)_{x \in V}$ with index $1/2$ admitting for kernel the Green function of a transient Markov chain on $V$, the sub-graph $\{[x,y] \in E : \varphi_x > h \text{ and } \varphi_y > h\}$ has a.s. an infinite connected component.
Proof of Theorem 2.1 Let \((\eta_x, x \in \mathcal{V})\) be a centered Gaussian process with covariance the Green function of a transient symmetric Markov chain on \(G\), \(U_0 = (U_n)_{n \geq 0}\). The law of \(U\) is characterized by its transition matrix \(P = (P(x,y))_{(x,y) \in \mathcal{V}^2}\). By assumption: \(P(x,y) = P(y,x), \forall x, y \in \mathcal{V}\). Denote by \(g = (g(x,y), (x, y) \in \mathcal{V}^2)\) its Green function i.e.

\[
g(x, y) = \sum_{n=0}^{\infty} P_x[U_n = y].
\]

Note that for every \(x\) in \(\mathcal{V}\): \(g(x, x) \geq 1\).

As usual, one can associate to \(U\) a time continuous Markov chain \(X = (X_t)_{t \geq 0}\) by spending at each site of \(\mathcal{V}\) an exponential time with parameter 1, independently of the jumps that are then performed according to \(P\) to leave the site. Then \(X\) is a transient symmetric Markov process with the same Green function as \(U\) and admitting a local time process \((L^x_t, x \in \mathcal{V}, t \geq 0)\).

Fix a finite subset \(J = \{x_1, x_2, \ldots, x_n\}\) of \(\mathcal{V}\) and set: \(G = (g(x_i, x_j))_{1 \leq i, j \leq n}\). We now compute an alternative expression of \(G\). To do so we reproduce an argument used in a particular case in [11] (proof of Theorem 2.1).

Set \(\sigma = \inf\{t \geq 0 : X_t \in \{x_1, x_2, \ldots, x_n\} \setminus \{X_0\}\}\).

The time \(\sigma\) may be infinite in that case the value of \(X_\sigma\) is a cemetery point. Set:

\[
b_{ij} = \mathbb{E}_{x_i}[L^x_\sigma] \quad \text{and} \quad q_{ij} = \mathbb{P}_{x_i}[X_\sigma = x_j].
\]

Note that: \(b_{ii} = 0\) for \(i \neq j\), \(q_{ii} = 0\) and \(\sum_{j=1}^{n} q_{ij} = 1 - \mathbb{P}_{x_i}[\sigma = \infty]\). Thanks to the Markov property we have:

\[
g(x_i, x_j) = \mathbb{E}_{x_i}[L^x_\infty] = \mathbb{E}_{x_i}[L^x_\sigma] + \mathbb{E}_{x_i}[\sigma < \infty; \mathbb{E}_{X_\sigma}[L^x_\infty]]
\]

\[
= b_{ij} + \sum_{k=1}^{n} q_{ik} g(x_k, x_i).
\]

Let \(B\) and \(Q\) be the matrices defined by: \(B = (b_{ij})_{1 \leq i, j \leq n}\) and \(Q = (q_{ij})_{1 \leq i, j \leq n}\). The above computation shows that: \(G = B + QG\), equivalently: \(B = (I - Q)G\).

Since \(X\) spends an exponential time with parameter 1 at each site before leaving it, we have: \(b_{ii} \geq 1\), for every \(i = 1, \ldots, n\). Consequently \(B\) is invertible and hence so are \((I - Q)\) and \(G\). One obtains:

\[
G = (I - Q)^{-1}B. \tag{2.1}
\]

For \(\lambda_1, \ldots, \lambda_n\) in \(\mathbb{R}_+\), denote by \(\lambda\) the diagonal matrix with entries \(\lambda_1, \ldots, \lambda_n\). Note that

\[
\det(I + \lambda G) = \det(I + \lambda (I - Q)^{-1}B) = \det(I + \lambda B^{1/2}(I - Q)^{-1}B^{1/2}),
\]

where \(B^{1/2}\) is the diagonal matrix with entries \(\sqrt{b_{11}}, \ldots, \sqrt{b_{nn}}\).

Consequently: \((|\eta_{x_i}|)_{1 \leq i \leq n}) \overset{(law)}{=} (|\phi_{x_i}|)_{1 \leq i \leq n}\), where \((\phi_{x_i})_{1 \leq i \leq n}\) is a centered Gaussian vector with covariance \(B^{1/2}(I - Q)^{-1}B^{1/2}\).

The centered Gaussian vector \((\frac{\phi_{x_i}}{\sqrt{b_{ii}}})_{1 \leq i \leq n}\) has \((I - Q)^{-1}\) for covariance.

Since: \((I - Q) \leq I\), one deduces, thanks to Theorem 2.3 in the paper of Karlin and Rinott [16], that:

\[
(|N_{x_i}|)_{1 \leq i \leq n} \sim (\frac{\phi_{x_i}}{\sqrt{b_{ii}}})_{1 \leq i \leq n}.
\]
where \((N_x)_{x \in V}\) is a family of i.i.d. standard Gaussian variables and the symbol “\(\prec\)”
denotes a relation of stochastic domination.

Remember that for every \(i = 1, \ldots, n\): \(b_{ii} \geq 1\), to conclude that

\[
(|N_x|)_{1 \leq i \leq n} \prec (|\phi_x|)_{1 \leq i \leq n},
\]

which can also be written:

\[
(|N_x|)_{x \in J} \prec (|\eta_x|)_{x \in J}.
\]

Since the above stochastic domination is true for every finite subset \(J\), by a simple
limiting argument, one obtains:

\[
(|N_x|)_{x \in V} \prec (|\eta_x|)_{x \in V}.
\] (2.2)

Consequently by Strassen Theorem, there exists a coupling \((|\tilde{\eta}_x|, |\tilde{N}_x|)\) of \(|\eta_x|\) and \(|N_x|\)
such that for every \(x \in V\): \(|\tilde{N}_x| \geq |\tilde{\eta}_x|\). One hence has for every \(h \geq 0\)
\(\{x \in V : |\tilde{N}_x| > h\} \subset \{x \in V : |\tilde{\eta}_x| > h\}\).

Let \(a\) be an element of \(V\). If \(P[|N_a| > h] > p_c^{s.}(G)\), then a.s. \(\{x \in V : |\tilde{N}_x| > h\}\) has
an infinite connected component and hence \(\{x \in V : |\tilde{\eta}_x| > h\}\) has also an infinite
connected component. □

**Proof of Theorem 2.2** Let \((\varphi_x, x \in V)\) be a permanental process with index 1/2
and kernel \((g(x, y), (x, y) \in V^2)\), the Green function of a transient random walk on \(V\).
This random walk is not assumed to be symmetric.

Fix a finite subset \(J = \{x_1, x_2, \ldots, x_n\}\) of \(V\) and set: \(G = (g(x_i, x_j))_{1 \leq i, j \leq n}\). The argument
developed in the proof of Theorem 2.1 to establish (2.1) does not require symmetry.
Keeping the same notation, one has: \(G = (I - Q)^{-1}B\), which leads to:

\[
\det(I + \lambda G) = \det(I + \lambda B^{1/2}(I - Q)^{-1}B^{1/2}).
\]

One obtains that way that \((\varphi_x)_{1 \leq i \leq n}\) admits for kernel the matrix \(B^{1/2}(I - Q)^{-1}B^{1/2}\).
Consequently the permanental vector \((\varphi_x / b_{ii})_{1 \leq i \leq n}\) admits for kernel \((I - Q)^{-1}\).

We use now an extension of the result of Karlin and Rinott (Theorem 2.1 in [16]) to
permanental vectors established by Marcus and Rosen [17]. They call it the permanen-
tal inequality. For simplicity, we enunciate it in the particular case we are interested in.

**Permanental inequality** Let \(\phi\) be a permanental vector with index 1/2 admitting
for kernel a non-singular matrix \(K = (K_{ij})_{1 \leq i, j \leq n}\). Assume that \(K^{-1} = I - Q\) with
\(Q_{ii} = 0\) for every \(i = 1, \ldots, n\). Then :

\[
(N_x^2)_{1 \leq i \leq n} \prec (\phi_i)_{1 \leq i \leq n},
\]

where the variables \(N_i, i = 1, \ldots, n,\) are i.i.d. centered real standard Gaussian variables.
One obtains:

\[(N^2_{x_i})_{1 \leq i \leq n} \prec (\frac{\varphi_{x_i}}{b_{ii}})_{1 \leq i \leq n},\]

where \((N_x)_{x \in V}\) is a family of i.i.d. standard Gaussian variables.

Since for every \(i = 1, \ldots, n\): \(b_{ii} \geq 1\), one has

\[(N^2_{x_i})_{1 \leq i \leq n} \prec (\varphi_{x_i})_{1 \leq i \leq n},\]

which leads to

\[(N^2_x)_{x \in V} \prec (\varphi_x)_{x \in V},\]

and hence to the conclusion. □

Remark 2.3 Actually Theorem 2.2 holds for any index \(\beta > 0\). More precisely, define \(h_{G,\beta}\) as the number such that:

\[P[\Gamma_\beta > h_{G,\beta}] = p^{\text{site}}(G)\]

where \(\Gamma_\beta\) is a gamma random variable with shape parameter \(\beta\) and scale parameter 1 (i.e. with density \(x^{\beta-1}e^{-x/\Gamma(\beta)}1_{x \geq 0}\)).

Then for every \(h < h_{G,\beta}\), and every permanental process \((\varphi_x, x \in V)\) with index \(\beta\) admitting for kernel the Green function of a transient random walk on \(G\): \(\{x \in V : \varphi_x > h\}\) has a.s. an infinite connected component. This is obtained similarly thanks to the general version of Marcus and Rosen’s permanental inequality [17].

3 Critical level sets for the Gaussian free fields

The graphs that we are considering are always non-oriented, locally bounded, infinite and connected. We assume that they are subsets of \(\mathbb{R}^d\) for some \(d \geq 1\). The next proposition extends the result of Bricmont et al [5] from the simple symmetric random walk on \(\mathbb{Z}^d\) to any transient simple random walk on an infinite graph. By simple random symmetric walk on a graph \(G = (V, E)\), one means a homogenous Markov chain such that its transition matrix \(P = (P(x, y))_{(x, y) \in V^2}\) satisfies:

\[P(x, y) = P(y, x),\]

\[P(x, y) > 0\] if \([x, y] \in E\) and \(P(x, y') = P(x, y'')\) for any \(y', y''\) such that \([x, y'], [x, y'']\) are in \(E\). The existence of such a process on \(G\) implies that \(G\) is regular in the sense that the vertices all have the same degree. One can hence assume without changing the matrix \(P\) that the edges all have the same length. Such a graph is said to be transient when the simple symmetric random walk on \(G\) is transient.

Given a graph \(G = (V, E)\) and a real valued process \((\eta_x, x \in V)\), to mean that the subgraph of \(G\) with vertex set \(\{x \in V : \eta_x > h\}\) and edge set \(\{[x, y] \in E : \eta_x > h\}\) and \(\eta_y > h\}\) contains an infinite connected subset of \(G\), we just write: \(\{\eta > h\}\) percolates.
Proposition 3.1 Let $G$ be a transient, regular, locally finite, infinite connected graph. Let $\{\eta_x \}_{x \in V}$ be a centered Gaussian field with covariance the Green function of the simple symmetric random walk on $G$. Set 

$$h^*_x(G) = \sup \{ h \in \mathbb{R} : \mathbb{P}[\{ \eta > h \} \text{ percolates}] > 0 \}$$

then one has:

$$h^*_x(G) \geq 0.$$

Proof Denote by $(g(x, y), (x, y) \in V^2)$ the Green function of the simple random walk on $G$. Let $(\eta_x, x \in V)$ be a centered Gaussian field with covariance $(g(x, y), (x, y) \in V^2)$. We extend $(\eta_x, x \in V)$ to the whole graph $G$. To do so we make use of Brownian motion on the graph $G$. We refer to Varopoulos [23], Chacon and Baxter [3], Barlow, Pitman and Yor [2] or Enriquez and Kifer [13], for various ways to construct $B$. This process can be roughly described as follows.

Starting from a point in the interior of an edge of $G$, $B$ moves along this edge as a real Brownian motion until it reaches one of its end points, call it $x_o$. At this time $B$ chooses uniformly an edge coming out of $x_o$, independently of the past, and moves like a real valued Brownian motion on the chosen edge until it reaches one of its end points.

Since the simple symmetric random walk on $G$ is transient, $B$ is transient. Denote by $\tilde{g} = (\tilde{g}(x, y), (x, y) \in G)$ the Green function of $B$. The restriction of $\tilde{g}$ to $V \times V$ coincides with $g$. Then define the extension $(\tilde{\eta}_x, x \in G)$ of $(\eta_x, x \in V)$ to $G$, by $E[\tilde{\eta}_x \tilde{\eta}_y] = \tilde{g}(x, y)$, for $x, y$ in $G$. One chooses $\tilde{\eta}$ independent of $B$.

One defines a distance $d$ on $G$ by using the Lebesgue distance, inside any given edge and by defining the distance between two vertices of $V$ as the minimal sum of length of edges necessary to connect one to the other.

The process $B$ admits a local time process $(L_x^t, x \in G, t \geq 0)$ which is continuous with respect to $d \times \text{Leb}(\mathbb{R}^+)$. Indeed, the local time process is obviously continuous at each point of $G \setminus V$. What about continuity at a point of $V$? Since it is a local question, it is equivalent to study the continuity at 0 of the local time process of a Brownian motion on a graph with one single vertex 0 and $n$ infinite edges coming out of 0, with equal probability to be chosen starting from 0. According to Theorem 2.1 (2.18) in [15], the local time process at 0 of this process is continuous. Consequently the local time process of $B$ is continuous on $G$. By Theorem 1 in [18], we hence know that $(\tilde{\eta}_x, x \in G)$ is continuous with respect to $d$.

Fix an element $a$ of $V$. Condition on $(B_0 = a)$, the laws of $(L_\infty^x, x \in G)$ and $(\tilde{\eta}_x, x \in G)$ can be connected thanks to a so-called isomorphism theorem established in [10], as follows:

$$(L_\infty^x + \frac{1}{2}(\tilde{\eta}_x + r)^2, x \in G) \overset{\text{(law)}}{=} \frac{1}{2}(\tilde{\eta}_x + r)^2, x \in G) \text{ under } E[\frac{\tilde{\eta}_a + r}{r}; \cdot] \quad (3.1)$$

for every non-zero real number $r$. 
The set \( \{ x \in G : L_x^\infty > 0 \} \) is a.s. an infinite connected subset of \( G \). It is the support of \( B \). Consequently, thanks to (3.1), one has for every \( r \neq 0 \):

\[
1 = \mathbb{E}\left[ \eta_x + \frac{r}{r} \middle| \{ x \in G : (\tilde{\eta}_x + r)^2 > 0 \} \right. \text{ has an infinite connected component}\]

which implies:

\[
\mathbb{P}\left[ \{ x \in G : (\tilde{\eta}_x + r)^2 > 0 \} \right. \text{ has an infinite connected component} \] > 0. \hfill (3.2)

Let \( C \) be an infinite connected subset of \( G \) such that : \( |\tilde{\eta}_x + r| > 0, \forall x \in C \). Then, because of the continuity of \( \tilde{\eta} \), either : \( \tilde{\eta}_x < -r, \forall x \in C \); either: \( \tilde{\eta}_x > -r, \forall x \in C \). Using (3.2) one obtains:

\[
\begin{align*}
\mathbb{P}\left[ \{ x \in G : \tilde{\eta}_x > -r \} \right. \text{ has an infinite connected component} \\
+ \mathbb{P}\left[ \{ x \in G : \tilde{\eta}_x < -r \} \right. \text{ has an infinite connected component} > 0.
\end{align*}
\]

and equivalently:

\[
\begin{align*}
\mathbb{P}\left[ \{ x \in G : \tilde{\eta}_x > -r \} \right. \text{ has an infinite connected component} \\
+ \mathbb{P}\left[ \{ x \in G : \tilde{\eta}_x > -r \} \right. \text{ has an infinite connected component} > 0,
\end{align*}
\]

which leads to: \( 2\mathbb{P}\left[ \{ x \in G : \tilde{\eta}_x > -|r| \} \right. \text{ has an infinite connected component} > 0, \)

for every \( r \neq 0 \). We finally obtain:

\[
\mathbb{P}\left[ \{ x \in \mathcal{V} : \eta_x > -|r| \} \right. \text{ has an infinite connected component} > 0 \text{ for every } r \neq 0. \quad \square
\]

To show Proposition 3.1 one could have used Sznitman’s interlacement isomorphism Theorem [22] instead of (3.1). The interest of (3.1) lies on the fact that it involves more elementary notions.

References

[1] Abächerli A. and Sznitman A.-S. : Level-set percolation for the Gaussian free field on a transient tree. Ann. Inst. Henri Poincar Probab. Stat. 54, no. 1, 173-201 (2018).

[2] Barlow M., Pitman J. and Yor M. : On Walsh’s Brownian motions. Séminaire de Probabilités, XXIII, 275-293, Lecture Notes in Math., 1372, Springer, Berlin (1989).

[3] Baxter J. R. and Chacon R V. : The equivalence of diffusions on networks to Brownian motion. Contemporary Mathematics 26, 33-48 (1984).

[4] Benjamini I. and Hermon J.: Recurrence of Markov chain traces. arXiv:1711.03479 (2019).
[5] Bricmont J., Lebowitz, J. L. and Maes, C.: Percolation in strongly correlated systems: the massless Gaussian field. J. Statist. Phys. 48 no. 5-6, 1249-1268 (1987).

[6] Burton R. M. and Waymire, E.: The central limit problem for infinitely divisible random measures. Dependence in probability and statistics (Oberwolfach, 1985), 383-395, Progr. Probab. Statist., 11, Birkhäuser Boston, Boston, MA (1986).

[7] Campanino M. and Russo L.: An upper bound on the critical percolation probability for the three-dimensional cubic lattice. Ann. Probab. 13, no. 2, 478-491 (1985).

[8] Drewitz A., Prévost A. and Rodriguez P.-F.: The sign clusters of the massless Gaussian free field percolate on $\mathbb{Z}^d$, $d \geq 3$ (and more). Comm. Math. Phys. 362, no. 2, 513-546 (2018).

[9] Drewitz A., Prévost A. and Rodriguez P.-F.: Geometry of Gaussian free field sign clusters and random interlacements. arXiv:1811.05970 (2018).

[10] Eisenbaum N.: Une version sans conditionnement du théorème d’isomorphisme de Dynkin. Séminaire de Probabilités, XXIX, 266-289, Lecture Notes in Math., 1613, Springer, Berlin (1995).

[11] Eisenbaum N.: On the infinite divisibility of squared Gaussian processes. Probab. Theory Related Fields 125, 381-392 (2003).

[12] Eisenbaum N. and Kaspi H.: On permanental processes. Stochastic Proc. Appl. 119, 5,1401-1415 (2009).

[13] Enriquez N. and Kifer Y.: Markov chains on graphs and Brownian motion. J. Theoret. Probab. 14, no. 2, 495-510 (2001).

[14] Higuchi Y.: Coexistence of the infinite (*) clusters: a remark on the square lattice site percolation. Z. Wahrsch. Verw. Gebiete 61, no. 1, 75-81 (1982).

[15] Ichiba T., Karatzas I., Prokaj V. and Yan M.: Stochastic integral equations for Walsh semimartingales. Ann. Inst. Henri Poincaré Probab. Stat. 54 no. 2, 726-756 (2018).

[16] Karlin S. and Rinott Y.: Total positivity properties of absolute value multinormal variables with applications to confidence interval estimates and related probabilistic inequalities. Ann. Statist. 9, no. 5,1035-1049 (1981).

[17] Marcus M. B. and Rosen J.: Conditions for permanental processes to be unbounded. Ann. Probab., 45,2059-2086 (2017).

[18] Marcus M. B. and Rosen J.: Sample path properties of the local times of strongly symmetric Markov processes via Gaussian processes. Ann. Probab. 20, no. 4, 1603-1684 (1992).
[19] Pitt L. D.: Positively correlated normal variables are associated. *Ann. Probab.* 10, no. 2, 496-499 (1982).

[20] Rodriguez P.-F. and Sznitman A.-S.: Phase transition and level-set percolation for the Gaussian free field. *Comm. Math. Phys.* 320, 571-601 (2013).

[21] Rodriguez P.-F.: Level set percolation for random interlacements and the Gaussian free field. *Stochastic Process. Appl.* 124, 1469-1502 (2014).

[22] Sznitman A.-S.: An isomorphism theorem for random interlacements. *Electron. Commun. Probab.* 17, no. 9, 9 pp. (2012).

[23] Varopoulos N. Th.: Long range estimates for Markov chains. *Bull. Sci. Math. (2)* 109, no. 3, 225-252 (1985).

Nathalie Eisenbaum
Laboratoire MAP5
Université de Paris
45, rue des Saints Pères
75270 Paris Cedex 06
nathalie.eisenbaum@parisdescartes.fr