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# A note on the extremal process of the supercritical Gaussian Free Field\*

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#### **Abstract**

We consider both the infinite-volume discrete Gaussian Free Field (DGFF) and the DGFF with zero boundary conditions outside a finite box in dimension larger or equal to 3. We show that the associated extremal process converges to a Poisson point process. The result follows from an application of the Stein-Chen method from [5].

**Keywords:** Extremal process; Gaussian free field; point processes; Poisson approximation; Stein-Chen method.

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## 1 Introduction

In this article we study the behavior of the extremal process of the DGFF in dimension larger or equal to 3. This extends the result presented in [9] in which the convergence of the rescaled maximum of the infinite-volume DGFF and the 0-boundary condition field was shown. It was proved there that the field belongs to the maximal domain of attraction of the Gumbel distribution; hence, a natural question that arises is that of describing more precisely its extremal points. In dimension 2, this was carried out by [6, 7] complementing a result of [8] on the convergence of the maximum; namely, the characterization of the limiting point process with a random mean measure yields as by-product an integral representation of the maximum. The extremes of the DGFF in dimension 2 have deep connections with those of Branching Brownian Motion ([1, 2, 3, 4]). These works showed that the limiting point process is a randomly shifted decorated Poisson point process, and we refer to [15] for structural details. In  $d \geq 3$ , one does not get a non-trivial decoration but instead a Poisson point process analogous to the extremal process of independent Gaussian random variables. To be more precise, we let  $E:=[0,\,1]^d\times(-\infty,\,+\infty]$  and  $V_N:=[0,\,n-1]^d\cap\mathbb{Z}^d$  the hypercube of volume  $N=n^d$ . Let  $(\varphi_{\alpha})_{\alpha\in\mathbb{Z}^d}$  be the infinite-volume DGFF, that is a centered Gaussian field on the square lattice with covariance  $g(\cdot, \cdot)$ , where g is the Green's function of the simple random walk. We define the following sequence of point processes on E:

$$\eta_n(\cdot) := \sum_{\alpha \in V_N} \varepsilon_{\left(\frac{\alpha}{n}, \frac{\varphi_\alpha - b_N}{a_N}\right)}(\cdot) \tag{1.1}$$

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where  $\varepsilon_x(\cdot)$ ,  $x \in E$ , is the point measure that gives mass one to a set containing x and zero otherwise, and

$$b_N := \sqrt{g(0)} \left[ \sqrt{2\log N} - \frac{\log\log N + \log(4\pi)}{2\sqrt{2\log N}} \right], \qquad a_N := g(0)(b_N)^{-1}.$$
 (1.2)

Here g(0) denotes the variance of the DGFF. Our main result is

**Theorem 1.1.** For the sequence of point processes  $\eta_n$  defined in (1.1) we have that

$$\eta_n \stackrel{d}{\to} \eta$$
,

as  $n \to +\infty$ , where  $\eta$  is a Poisson random measure on E with intensity measure given by  $\mathrm{d}\, t \otimes (\mathrm{e}^{-z}\, \mathrm{d}\, z)$  where  $\mathrm{d}\, t \otimes \mathrm{d}\, z$  is the Lebesgue measure on E, and  $\stackrel{d}{\to}$  is the convergence in distribution on  $\mathcal{M}_p(E)^1$ .

The proof is based on the application of the two-moment method of [5] that allows us to compare the extremal process of the DGFF and a Poisson point process with the same mean measure. To prove that the two processes converge, we will exploit a classical theorem by Kallenberg.

It is natural then to consider also convergence for the DGFF  $(\psi_{\alpha})_{\alpha \in \mathbb{Z}^d}$  with zero boundary conditions outside  $V_N$ . For the sequences of point measures

$$\rho_n(\cdot) := \sum_{\alpha \in V_N} \varepsilon_{\left(\frac{\alpha}{n}, \frac{\psi_\alpha - b_N}{a_N}\right)}(\cdot) \tag{1.3}$$

we establish the following Theorem:

**Theorem 1.2.** For the sequence of point processes  $\rho_n$  defined in (1.3) we have that

$$\rho_n \stackrel{d}{\to} \eta$$
,

as  $n \to +\infty$  in  $\mathcal{M}_p(E)$ , where  $\eta$  is as in Theorem 1.1.

The convergence is shown by reducing ourselves to check the conditions of Kallenberg's Theorem on the bulk of  $V_N$ , where we have a good control on the drift of the conditioned field, and then by showing that the process on the whole of  $V_N$  and on the bulk are close as n becomes large.

The outline of the paper is as follows. In Section 2 we will recall the definition of DGFF and the Stein-Chen method, while Section 3 and Section 4 are devoted to the proofs of Theorems 1.1 and 1.2 respectively.

## 2 Preliminaries

## 2.1 The DGFF

Let  $d \geq 3$  and denote with  $\|\cdot\|$  the  $\ell_\infty$ -norm on  $\mathbb{Z}^d$ . Let  $\psi = (\psi_\alpha)_{\alpha \in \mathbb{Z}^d}$  be a discrete Gaussian Free Field with zero boundary conditions outside  $\Lambda \subset \mathbb{Z}^d$ . On the space  $\Omega := \mathbb{R}^{\mathbb{Z}^d}$  endowed with its product topology, its law  $\widetilde{\mathsf{P}}_\Lambda$  can be explicitly written as

$$\widetilde{\mathsf{P}}_{\Lambda}(\mathrm{d}\,\psi) = \frac{1}{Z_{\Lambda}} \exp\left(-\frac{1}{4d} \sum_{\alpha,\,\beta \in \mathbb{Z}^d:\, \|\alpha - \beta\| = 1} (\psi_{\alpha} - \psi_{\beta})^2\right) \prod_{\alpha \in \Lambda} \mathrm{d}\,\psi_{\alpha} \prod_{\alpha \in \mathbb{Z}^d \setminus \Lambda} \varepsilon_0(\psi_{\alpha}).$$

In other words  $\psi_{\alpha}=0$   $\widetilde{\mathsf{P}}_{\Lambda}$ -a. s. if  $\alpha\in\mathbb{Z}^d\setminus\Lambda$ , and  $(\psi_{\alpha})_{\alpha\in\Lambda}$  is a multivariate Gaussian random variable with mean zero and covariance  $(g_{\Lambda}(\alpha,\beta))_{\alpha,\,\beta\in\mathbb{Z}^d}$ , where  $g_{\Lambda}$  is the Green's

 $<sup>{}^1\</sup>mathcal{M}_p(E)$  denotes the set of (Radon) point measures on E endowed with the topology of vague convergence.

function of the discrete Laplacian problem with Dirichlet boundary conditions outside  $\Lambda$ . For a thorough review on the model the reader can refer for example to [16]. It is known [10, Chapter 13] that the finite-volume measure  $\psi$  admits an infinite-volume limit as  $\Lambda \uparrow \mathbb{Z}^d$  in the weak topology of probability measures. This field will be denoted as  $\varphi = (\varphi_\alpha)_{\alpha \in \mathbb{Z}^d}$ . It is a centered Gaussian field with covariance matrix  $g(\alpha, \beta)$  for  $\alpha, \beta \in \mathbb{Z}^d$ . With a slight abuse of notation, we write  $g(\alpha - \beta)$  for  $g(0, \alpha - \beta)$  and also  $g_\Lambda(\alpha) = g_\Lambda(\alpha, \alpha)$ . g admits a so-called random walk representation: if  $\mathbb{P}_\alpha$  denotes the law of a simple random walk S started at  $\alpha \in \mathbb{Z}^d$ , then

$$g(\alpha, \beta) = \mathbb{E}_{\alpha} \left[ \sum_{n \geq 0} \mathbb{1}_{\{S_n = \beta\}} \right].$$

In particular this gives  $g(0) < +\infty$  for  $d \ge 3$ . A comparison of the covariances in the infinite and finite-volume is possible in the *bulk* of  $V_N$ : for  $\delta > 0$  this is defined as

$$V_N^{\delta} := \left\{ \alpha \in V_N : \|\alpha - \beta\| > \delta n, \, \forall \, \beta \in \mathbb{Z}^d \setminus V_N \right\}. \tag{2.1}$$

In order to compare covariances in the finite and infinite-volume field, we recall the following Lemma, whose proof is presented in [9, Lemma 7]).

**Lemma 2.1.** For any  $\delta > 0$  and  $\alpha, \beta \in V_N^{\delta}$  one has

$$g(\alpha, \beta) - C_d \left(\delta N^{1/d}\right)^{2-d} \le g_{V_N}(\alpha, \beta) \le g(\alpha, \beta).$$
 (2.2)

In particular we have,  $g_{V_N}(\alpha)=g(0)\left(1+\mathrm{O}\left(N^{(2-d)/d}\right)\right)$  uniformly for  $\alpha\in V_N^\delta$ .

### 2.2 The Stein-Chen method

As main tool of this article we will use (and restate here) a theorem from [5]. Consider a sequence of Bernoulli random variables  $(X_{\alpha})_{\alpha \in \mathcal{I}}$  where  $X_{\alpha} \sim Be(p_{\alpha})$  and  $\mathcal{I}$  is some index set. For each  $\alpha$  we define a subset  $B_{\alpha} \subseteq \mathcal{I}$  which we consider a "neighborhood" of dependence for the variable  $X_{\alpha}$ , such that  $X_{\alpha}$  is nearly independent from  $X_{\beta}$  if  $\beta \in \mathcal{I} \setminus B_{\alpha}$ . Set

$$\begin{split} b_1 := \sum_{\alpha \in \mathcal{I}} \sum_{\beta \in B_{\alpha}} p_{\alpha} p_{\beta}, \\ b_2 := \sum_{\alpha \in \mathcal{I}} \sum_{\alpha \neq \beta \in B_{\alpha}} \mathsf{E} \left[ X_{\alpha} X_{\beta} \right], \\ b_3 := \sum_{\alpha \in \mathcal{I}} \mathsf{E} \left[ \left| \mathsf{E} \left[ X_{\alpha} - p_{\alpha} \mid \mathcal{H}_1 \right] \right| \right] \end{split}$$

where

$$\mathcal{H}_1 := \sigma(X_\beta : \beta \in \mathcal{I} \setminus B_\alpha)$$
.

**Theorem 2.2** ([5, Theorem 2]). Let  $\mathcal{I}$  be an index set. Partition the index set  $\mathcal{I}$  into disjoint non-empty sets  $\mathcal{I}_1, \ldots, \mathcal{I}_k$ . For any  $\alpha \in \mathcal{I}$ , let  $(X_\alpha)_{\alpha \in \mathcal{I}}$  be a dependent Bernoulli process with parameter  $p_\alpha$ . Let  $(Y_\alpha)_{\alpha \in \mathcal{I}}$  be independent Poisson random variables with intensity  $p_\alpha$ . Also let

$$W_j := \sum_{\alpha \in \mathcal{I}_j} X_\alpha \quad \text{ and } \quad Z_j := \sum_{\alpha \in \mathcal{I}_j} Y_\alpha \quad \text{ and } \quad \lambda_j := \mathsf{E}[W_j] = \mathsf{E}[Z_j].$$

Then

$$\|\mathcal{L}(W_1,\ldots,W_k) - \mathcal{L}(Z_1,\ldots,Z_k)\|_{TV} \le 2\min\left\{1, \ 1.4\left(\min_{1\le j\le k}\lambda_j\right)^{-1/2}\right\} (2b_1 + 2b_2 + b_3)$$
 (2.3)

where  $\|\cdot\|_{TV}$  denotes the total variation distance and  $\mathcal{L}(W_1,\ldots,W_k)$  denotes the joint law of these random variables.

## 3 Proof of Theorem 1.1: the infinite-volume case

*Proof.* We recall that  $E = [0,1]^d \times (-\infty,+\infty]$  and  $V_N = [0,n-1]^d \cap \mathbb{Z}^d$ . To show the convergence of  $\eta_n$  to  $\eta$ , we will exploit Kallenberg's theorem [11, Theorem 4.7]. According to it, we need to verify the following conditions:

i) for any A, a bounded rectangle<sup>2</sup> in  $[0,1]^d$ , and  $(x,y] \subset (-\infty,+\infty]$ 

$$\mathsf{E}[\eta_n(A\times(x,y])]\to\mathsf{E}[\eta(A\times(x,y])]=|A|(\mathrm{e}^{-x}-\mathrm{e}^{-y}).$$

We adopt the convention  $e^{-\infty} = 0$  and the notation |A| for the Lebesgue measure of A.

ii) For all  $k \geq 1$ , and  $A_1, A_2, \ldots, A_k$  disjoint rectangles in  $[0,1]^d$  and  $R_1, R_2, \ldots, R_k$ , each of which is a finite union of disjoint intervals of the type  $(x, y] \subset (-\infty, +\infty]$ ,

$$P(\eta_n(A_1 \times R_1) = 0, ..., \eta_n(A_k \times R_k) = 0)$$

$$\to P(\eta(A_1 \times R_1) = 0, ..., \eta(A_k \times R_k) = 0) = \exp\left(-\sum_{j=1}^k |A_j| \omega(R_j)\right)$$
 (3.1)

where  $\omega(\mathrm{d}\,z) := \mathrm{e}^{-z}\,\mathrm{d}\,z$ .

Let us denote by  $u_N(z) := a_N z + b_N$ . The first condition follows by Mills ratio

$$\left(1 - \frac{1}{t^2}\right) \frac{e^{-t^2/2}}{\sqrt{2\pi}t} \le P\left(\mathcal{N}(0, 1) > t\right) \le \frac{e^{-t^2/2}}{\sqrt{2\pi}t}, \quad t > 0.$$
(3.2)

More precisely

$$\begin{split} \mathsf{E}[\eta_{n}(A\times(x,y])] &= \sum_{\alpha\in nA\cap V_{N}} \mathsf{P}\left(\varphi_{\alpha}\in(u_{N}(x),u_{N}(y)]\right) \\ &\leq \sum_{\alpha\in nA\cap V_{N}} \left(\frac{\mathrm{e}^{-\frac{u_{N}(x)^{2}}{2g(0)}}}{\sqrt{2\pi}u_{N}(x)} - \frac{\mathrm{e}^{-\frac{u_{N}(y)^{2}}{2g(0)}}}{\sqrt{2\pi}u_{N}(y)} \left(1 - \frac{1}{u_{N}(y)^{2}}\right)\right) \\ &\leq |nA\cap V_{N}| \left(\frac{\mathrm{e}^{-x+\mathrm{o}(1)}}{N} - \frac{\mathrm{e}^{-y+\mathrm{o}(1)}}{N} \left(1 - \frac{1}{2g(0)\log N(1+\mathrm{o}(1))}\right)\right) \\ &\to |A|(\mathrm{e}^{-x} - \mathrm{e}^{-y}). \end{split} \tag{3.4}$$

Similarly, one can plug in (3.3) the reverse bounds of (3.2) to prove the lower bound, and thus condition i).

To show ii), we need a few more details. Let  $k \geq 1$ ,  $A_1, \ldots, A_k$  and  $R_1, \ldots, R_k$  be as in the assumptions. Let us denote by  $\mathcal{I}_j = nA_j \cap V_N$  and  $\mathcal{I} = \mathcal{I}_1 \cup \ldots \cup \mathcal{I}_k$ . For  $\alpha \in \mathcal{I}_j$  define

$$X_{\alpha} := \mathbb{1}_{\left\{\frac{\varphi_{\alpha} - b_{N}}{a_{N}} \in R_{j}\right\}}$$

and  $p_{\alpha} := \mathsf{P}\left((\varphi_{\alpha} - b_N)/a_N \in R_j\right)$ . Choose now a small  $\epsilon > 0$  and fix the neighborhood of dependence  $B_{\alpha} := B\left(\alpha, (\log N)^{2+2\epsilon}\right) \cap \mathcal{I}^3$  for  $\alpha \in \mathcal{I}$ . Let  $W_j := \sum_{\alpha \in \mathcal{I}_j} X_{\alpha}$  and  $Z_j$  be as in Theorem 2.2.

By the simple observation that

$$P(\eta_n(A_1 \times R_1) = 0, ..., \eta_n(A_k \times R_k) = 0) = P(W_1 = 0, ..., W_k = 0),$$

 $<sup>^2</sup>$ A bounded rectangle has the form  $J_1 \times \cdots \times J_d$  with  $J_i = [0, 1] \cap (a_i, b_i]$ ,  $a_i, b_i \in \mathbb{R}$  for all  $1 \le i \le d$ .  $^3B(x,r)$  denotes a ball of radius r centered at x

to prove the convergence (3.1), we can use Theorem 2.2 and show that the error bound on the RHS of (2.3) goes to 0.

First we bound  $b_1$  as follows. By definition of  $R_1, R_2, \ldots, R_k$ , there exists  $z \in \mathbb{R}$  such that  $R_j \subset (z, +\infty]$  for  $1 \le j \le k$ . Hence for any  $1 \le j \le k$ , for any  $\alpha \in \mathcal{I}_j$  we have that

$$p_{\alpha} = \mathsf{P}\left(\frac{\varphi_{\alpha} - b_{N}}{a_{N}} \in R_{j}\right) \le \mathsf{P}(\varphi_{\alpha} > u_{N}(z)) \overset{\text{(3.2)}}{\le} \frac{\mathrm{e}^{-\frac{u_{N}(z)^{2}}{2g(0)}}}{\sqrt{2\pi}u_{N}(z)} \sqrt{g(0)}.$$

The bound is independent of  $\alpha$  and j, therefore for some C > 0

$$b_1 \le CN(\log N)^{d(2+2\epsilon)} e^{-2z} N^{-2} \to 0.$$
 (3.5)

For  $b_2$  note that it was shown in [9] that for  $z \in \mathbb{R}$  and  $\alpha \neq \beta \in V_N$ 

$$\mathsf{P}(\varphi_{\alpha} > u_{N}(z), \, \varphi_{\beta} > u_{N}(z)) \leq \frac{(2-\kappa)^{3/2}}{\kappa^{1/2}} N^{-2/(2-\kappa)} \max \left\{ e^{-2z} \, \mathbb{1}_{\{z \leq 0\}}, e^{-2z/(2-\kappa)} \, \mathbb{1}_{\{z > 0\}} \right\}. \tag{3.6}$$

Here we have introduced  $\kappa:=\mathbb{P}_0\left(\widetilde{H}_0=+\infty\right)\in(0,1)$  and  $\widetilde{H}_0=\inf\{n\geq 1:\,S_n=0\}.$  Observe that for any  $1\leq j\leq k$ ,  $\alpha\in\mathcal{I}$  and  $\beta\in B_\alpha$  one has

$$\mathsf{E}[X_{\alpha}X_{\beta}] \le \mathsf{P}(\varphi_{\alpha} > u_N(z), \varphi_{\beta} > u_N(z))$$

so that by (3.6) we can find some constant  $C^\prime>0$  such that

$$b_2 \le C' N^{-\kappa/(2-\kappa)} (\log N)^{d(2+2\epsilon)} \max \left\{ e^{-2z} \, \mathbb{1}_{\{z \le 0\}}, e^{-2z/(2-\kappa)} \, \mathbb{1}_{\{z > 0\}} \right\} \to 0.$$

The error is similar to the estimate obtained in [9, Equation (8)]. Finally we need to handle  $b_3$ . From Section 2.2 we set for  $\alpha \in \mathcal{I}$ ,  $\mathcal{H}_1 := \sigma(X_\beta : \beta \in \mathcal{I} \setminus B_\alpha)$  and we define  $\mathcal{H}_2 := \sigma(\varphi_\beta : \beta \in \mathcal{I} \setminus B_\alpha)$ . We observe that

$$b_3 = \sum_{\alpha \in \mathcal{I}} \mathsf{E}\left[\left|\mathsf{E}\left[X_\alpha - p_\alpha \mid \mathcal{H}_1\right]\right|\right] \leq \sum_{\alpha \in \mathcal{I}} \mathsf{E}\left[\left|\mathsf{E}\left[X_\alpha \mid \mathcal{H}_2\right] - p_\alpha\right|\right]$$

since  $\mathcal{H}_1 \subseteq \mathcal{H}_2$  and using the tower property of the conditional expectation. Now denote by  $U_{\alpha} := \mathbb{Z}^d \setminus (\mathcal{I} \setminus B_{\alpha})$ . Let us abbreviate  $u_N(R_j) := \{u_N(y) : y \in R_j\}$ . Then for  $\alpha \in I_j$  and  $1 \leq j \leq k$ , by the Markov property of the DGFF [14, Lemma 1.2] we have that

$$\mathsf{E}\left[X_{\alpha} \mid \mathcal{H}_{2}\right] = \widetilde{\mathsf{P}}_{U_{\alpha}}(\psi_{\alpha} + \mu_{\alpha} \in u_{N}(R_{j})) \qquad \mathsf{P} - a. \ s.$$

where  $(\psi_{\alpha})_{\alpha \in \mathbb{Z}^d}$  is a Gaussian Free Field with zero boundary conditions outside  $U_{\alpha}$  and

$$\mu_{\alpha} = \sum_{\beta \in \mathcal{I} \setminus B_{\alpha}} \mathbb{P}_{\alpha} \left( H_{\mathcal{I} \setminus B_{\alpha}} < +\infty, \, S_{H_{\mathcal{I} \setminus B_{\alpha}}} = \beta \right) \varphi_{\beta}.$$

Here  $H_{\Lambda} := \inf \{ n \geq 0 : S_n \in \Lambda \}$ ,  $\Lambda \subset \mathbb{Z}^d$ . Now as in [9, Equation (10)] one can show, using the Markov property, that

$$\operatorname{Var}\left[\mu_{\alpha}\right] \leq \sup_{\beta \in \mathcal{I} \backslash B_{\alpha}} g(\alpha, \beta) \leq \frac{c}{(\log N)^{2(1+\epsilon)(d-2)}}$$

for some c > 0. Hence we get that there exists a constant c' > 0 (independent of  $\alpha$  and j) such that

$$\mathsf{P}\left(|\mu_{\alpha}| > (u_N(z))^{-1-\epsilon}\right) \le c' \exp\left(-(\log N)^{(2d-5)(1+\epsilon)}\right). \tag{3.7}$$

Recalling that  $R_j \subset (z, +\infty]$  for all  $1 \leq j \leq k$ , this immediately shows that for  $d \geq 3$ 

$$\sum_{j=1}^k \sum_{\alpha \in \mathcal{I}_j} \mathsf{E}\left[\left|\widetilde{\mathsf{P}}_{U_\alpha}(\psi_\alpha + \mu_\alpha \in u_N(R_j)) - p_\alpha\right| \mathbb{1}_{\left\{|\mu_\alpha| > (u_N(z))^{-1-\epsilon}\right\}}\right] \to 0.$$

So to show that  $b_3 \to 0$  we are left with proving

$$\sum_{j=1}^{k} \sum_{\alpha \in \mathcal{I}_{j}} \mathsf{E}\left[\left|\widetilde{\mathsf{P}}_{U_{\alpha}}(\psi_{\alpha} + \mu_{\alpha} \in u_{N}(R_{j})) - p_{\alpha}\right| \mathbb{1}_{\left\{|\mu_{\alpha}| \leq (u_{N}(z))^{-1-\epsilon}\right\}}\right] \to 0. \tag{3.8}$$

We now focus on the term inside the summation. For this, first we write  $R_j = \bigcup_{l=1}^m (w_l, r_l]$  with  $-\infty < w_1 < r_1 < w_2 < \cdots < r_m \le +\infty$  for some  $m \ge 1$ . Hence, we can expand the difference in the absolute value of (3.8) as follows:

$$\left(p_{\alpha} - \widetilde{\mathsf{P}}_{U_{\alpha}}(\psi_{\alpha} + \mu_{\alpha} \in u_{N}(R_{j}))\right) \\
= \sum_{l=1}^{m} \left(\mathsf{P}(\varphi_{\alpha} \in (u_{N}(w_{l}), u_{N}(r_{l})]) - \widetilde{\mathsf{P}}_{U_{\alpha}}(\psi_{\alpha} + \mu_{\alpha} \in (u_{N}(w_{l}), u_{N}(r_{l})])\right) \\
= \sum_{l=1}^{m} \left(\mathsf{P}(\varphi_{\alpha} > u_{N}(w_{l})) - \widetilde{\mathsf{P}}_{U_{\alpha}}(\psi_{\alpha} + \mu_{\alpha} > u_{N}(w_{l}))\right) \\
- \sum_{l=1}^{m} \left(\mathsf{P}(\varphi_{\alpha} > u_{N}(r_{l})) - \widetilde{\mathsf{P}}_{U_{\alpha}}(\psi_{\alpha} + \mu_{\alpha} > u_{N}(r_{l}))\right) \tag{3.9}$$

(if  $r_l = +\infty$  for some l, we conventionally set  $P(\varphi_\alpha > u_N(r_l)) = 0$  and similarly for the other summand). Using the triangular inequality in (3.8), it turns out that to finish it is enough to show that for an arbitrary  $w \in \mathbb{R}$ ,

$$\sum_{\alpha \in \mathcal{I}} \mathsf{E}\left[\left|\widetilde{\mathsf{P}}_{U_{\alpha}}(\psi_{\alpha} + \mu_{\alpha} > u_{N}(w)) - \mathsf{P}(\varphi_{\alpha} > u_{N}(w))\right| \mathbb{1}_{\left\{|\mu_{\alpha}| \leq (u_{N}(z))^{-1-\epsilon}\right\}}\right] \to 0. \tag{3.10}$$

For this, first we show that on  $\mathcal{Q}:=\left\{\mathsf{P}(\varphi_{\alpha}>u_{N}(w))>\widetilde{\mathsf{P}}_{U_{\alpha}}(\psi_{\alpha}+\mu_{\alpha}>u_{N}(w))\right\}$ 

$$T_{1,2} = \sum_{\alpha \in \mathcal{I}} \mathsf{E}\left[\left(\mathsf{P}(\varphi_{\alpha} > u_N(w)) - \widetilde{\mathsf{P}}_{U_{\alpha}}(\psi_{\alpha} + \mu_{\alpha} > u_N(w))\right) \mathbb{1}_{\left\{|\mu_{\alpha}| \leq (u_N(z))^{-1-\epsilon}\right\}} \mathbb{1}_{\mathcal{Q}}\right] \to 0. \tag{3.11}$$

This follows from the same estimates of  $T_{1,2}$  and Claim 6 of [9]. Indeed on  $\mathcal{Q}\cap\left\{|\mu_{\alpha}|\leq \left(u_{N}(z)\right)^{-1-\epsilon}\right\}$ 

$$\begin{split} & \sum_{\alpha \in \mathcal{I}} \left( \mathsf{P}(\varphi_{\alpha} > u_{N}(w)) - \widetilde{\mathsf{P}}_{U_{\alpha}} \left( \psi_{\alpha} + \mu_{\alpha} > u_{N}(w) \right) \right) \\ & \leq \sum_{\alpha \in \mathcal{I}} \frac{\sqrt{g(0)} \, \mathrm{e}^{-\frac{u_{N}(w)^{2}}{2g(0)}}}{\sqrt{2\pi} u_{N}(w)} \left( 1 - (1 + \mathrm{o}\,(1)) \left( \frac{\sqrt{g_{U_{\alpha}}(\alpha)} u_{N}(w) \, \mathrm{e}^{\left( 1 - \frac{g(0)}{g_{U_{\alpha}}(\alpha)} \right) \frac{u_{N}(w)^{2}}{2g(0)} + \mathrm{o}\,(1)}}{\sqrt{g(0)} u_{N}(w) (1 + \mathrm{o}\,(1))} \right) \right) \\ & \leq C N \frac{\sqrt{g(0)} \, \mathrm{e}^{-\frac{u_{N}(w)^{2}}{2g(0)}}}{\sqrt{2\pi} u_{N}(w)} \, \mathrm{o}\,(1) = \mathrm{o}\,(1) \,. \end{split}$$

Similarly one can show that on the complementary event  $Q^c$  (recall (3.11) for the definition of Q)

$$T_{1,1} = \sum_{\alpha \in \mathcal{I}} \mathsf{E}\left[\left(\widetilde{\mathsf{P}}_{U_\alpha}(\psi_\alpha + \mu_\alpha > u_N(w)) - \mathsf{P}(\varphi_\alpha > u_N(w))\right)\mathbbm{1}_{\left\{|\mu_\alpha| \leq (u_N(z))^{-1-\epsilon}\right\}} \mathbbm{1}_{\mathcal{Q}^c}\right] = \mathrm{o}\left(1\right).$$

This shows that  $b_3 \to 0$ . Hence from Theorem 2.2 it follows that

$$\left| \mathsf{P}(W_1 = 0, \dots, W_k = 0) - \prod_{j=1}^k \mathsf{P}(Z_j = 0) \right| = \mathrm{o}(1),$$

having used the independence of the  $Z_j$ 's. Notice that by definition  $Z_j$  is a Poisson random variable with intensity  $\sum_{\alpha \in \mathcal{I}_j} \mathsf{P}\left((\varphi_\alpha - b_N)/a_N \in R_j\right)$ . Decomposing  $R_j$  as a union of finite intervals and using Mills ratio, similarly to the argument leading to (3.4), one has

$$P(Z_i = 0) \rightarrow \exp(-|A_i|\omega(R_i))$$

(recall  $\omega(R_j) = \int_{R_j} \mathrm{e}^{-z} \, \mathrm{d} \, z$ ). Hence it follows that

$$\prod_{j=1}^{k} \mathsf{P}(Z_j = 0) \to \exp\left(-\sum_{j=1}^{k} |A_j| \omega(R_j)\right),\tag{3.12}$$

which completes the proof of ii) and therefore of Theorem 1.1.

### 4 Proof of Theorem 1.2: the finite-volume case

We will now show the theorem for the field with zero boundary conditions. As remarked in the Introduction, since on the bulk defined in (2.1) we have a good control on the conditioned field, we will first prove convergence therein, and then we will use a converging-together theorem to achieve the final limit. We will first need some notation used throughout the Section: first, we consider  $(\psi_{\alpha})_{\alpha \in V_N}$  with law  $\widetilde{\mathsf{P}}_N := \widetilde{\mathsf{P}}_{V_N}$ . We also use the shortcut  $g_N(\cdot,\cdot) = g_{V_N}(\cdot,\cdot)$ . We will need the notation  $\mathcal{C}_K^+(E)$  for the set of positive, continuous and compactly supported functions on  $E = [0,1]^d \times (-\infty,+\infty]$ . We first begin with a lemma on the point process convergence on bulk. Define a point process on E by

$$\rho_n^{\delta}(\cdot) = \sum_{\alpha \in V_N^{\delta}} \varepsilon_{\left(\frac{\alpha}{n}, \frac{\psi_{\alpha} - b_N}{a_N}\right)}(\cdot). \tag{4.1}$$

**Lemma 4.1.** Let  $\delta > 0$ . On  $\mathcal{M}_p(E)$ ,  $\rho_n^{\delta} \xrightarrow{d} \rho^{\delta}$  where  $\rho^{\delta}$  is a Poisson random measure with intensity  $\mathrm{d}\,t_{|_{[\delta,1-\delta]^d}} \otimes (\mathrm{e}^{-x}\,\mathrm{d}\,x)^4$ .

*Proof.* We will show i) and ii) of Page 4 (and from which we will borrow the notation starting from now).

i) We begin with an upper bound on  $\widetilde{\mathsf{E}}_N\left[\rho_n^\delta(A\times(x,y])\right]$ :

$$\sum_{\alpha \in nA \cap V_N^{\delta}} \widetilde{\mathsf{P}}_N(\psi_{\alpha} > u_N(x)) - \widetilde{\mathsf{P}}_N(\psi_{\alpha} > u_N(y))$$

$$\stackrel{(3.2)}{\leq} \sum_{\alpha \in nA \cap V_N^{\delta}} \frac{e^{-\frac{u_N(x)^2}{2g_N(\alpha)}}}{\sqrt{2\pi}u_N(x)} \sqrt{g_N(\alpha)} - \frac{e^{-\frac{u_N(y)^2}{2g_N(\alpha)}}}{\sqrt{2\pi}u_N(y)} \sqrt{g_N(\alpha)} (1 + o(1))$$

$$\stackrel{\text{Lemma 2.1}}{=} \sum_{\alpha \in nA \cap V_N^{\delta}} \frac{e^{-\frac{u_N(x)^2}{2g(0)(1+c_n)}}}{\sqrt{2\pi}u_N(x)} \sqrt{g(0)} (1 + c_n) - \frac{e^{-\frac{u_N(y)^2}{2g(0)(1+c_n)}}}{\sqrt{2\pi}u_N(y)} \sqrt{g(0)} (1 + c_n)$$

$$\stackrel{n \to +\infty}{=} (e^{-x} - e^{-y}) |A \cap [\delta, 1 - \delta]^d|. \tag{4.2}$$

We stress that in the second step the error term  $c_n := O\left(n^{2-d}\right)$  coming from Lemma 2.1 guarantees the convergence in the last line. The lower bound follows similarly. ii) To show the second condition we again use Theorem 2.2. Let  $A_1, \ldots, A_k$  and  $R_1, \ldots, R_k$ 

be as in proof of Theorem 1.1. Let  $\mathcal{I}_j := nA_j \cap V_N^{\delta}$  and  $\mathcal{I} = \mathcal{I}_1 \cup \cdots \cup \mathcal{I}_k$ . For  $\epsilon > 0$  we are setting  $B_{\alpha} := B\left(\alpha, \, (\log N)^{2(1+\epsilon)}\right) \cap \mathcal{I}$ . Note that, albeit slightly different, we are using the same notations for the neighborhood of dependence and the index sets of Section 3,

 $<sup>^4\</sup>mathrm{d}\,t_{|_{[\delta,1-\delta]^d}} \text{ is the restriction of the Lebesgue measure to } [\delta,1-\delta]^d.$ 

but no confusion should arise. Observe that there exists  $z \in \mathbb{R}$  such that for all  $1 \le j \le k$ ,  $R_j \subset (z, \infty]$ ; we have

$$p_{\alpha} = \widetilde{\mathsf{P}}_{N} \left( \frac{\psi_{\alpha} - b_{N}}{a_{N}} \in u_{N}(R_{j}) \right) \leq \widetilde{\mathsf{P}}_{N} \left( \psi_{\alpha} > u_{N}(z) \right) \overset{\text{(3.2)}}{\leq} \frac{\mathrm{e}^{-\frac{u_{N}(z)^{2}}{2g(0)}}}{\sqrt{2\pi}u_{N}(z)} \sqrt{g(0)}$$

where we have also used the fact that  $g_N(\alpha) \leq g(0)$ . The bound on  $b_1$  (cf. Theorem 2.2) follows exactly as in (3.5) and yields that, for some C > 0,

$$b_1 \le CN(\log N)^{d(2+2\epsilon)} e^{-2z} N^{-2} \to 0.$$

The calculation of  $b_2$  can be performed similarly using the covariance matrix of the vector  $(\psi_{\alpha},\,\psi_{\beta})$ ,  $\alpha \neq \beta \in V_N^{\delta}$  and Lemma 2.1. This gives that for some  $C,\,C'>0$  independent of  $\alpha,\beta \in V_N^{\delta}$ 

$$b_2 \le \sum_{\alpha \in \mathcal{I}} \sum_{\beta \in B_{\alpha}} \frac{C}{\log N} \exp\left(-\frac{u_N(z)^2}{g(0) + g(\alpha - \beta)} \left(1 + O\left(N^{(2-d)/d}\right)\right)\right)$$

$$\le C' N^{-\kappa/(2-\kappa)} (\log N)^{2d(1+\epsilon)} \max\left\{e^{-2z} \, \mathbb{1}_{\{z \le 0\}}, \, e^{-2z/(2-\kappa)} \, \mathbb{1}_{\{z > 0\}}\right\} \to 0$$

(cf. [9, Equation (8)]). Note the estimate for  $b_2$  is exactly same as in the infinite volume case.

We will now pass to  $b_3$ . We repeat our choice of  $\mathcal{H}_1 = \sigma(X_\beta : \beta \in \mathcal{I} \setminus B_\alpha)$  and  $\mathcal{H}_2 = \sigma(\psi_\beta : \beta \in \mathcal{I} \setminus B_\alpha)$  so that  $b_3$  becomes

$$\sum_{j=1}^{k} \sum_{\alpha \in \mathcal{I}_{j}} \widetilde{\mathsf{E}}_{N} \left[ \left| \widetilde{\mathsf{E}}_{N} \left[ X_{\alpha} - p_{\alpha} | \mathcal{H}_{1} \right] \right| \right] \leq \sum_{j=1}^{k} \sum_{\alpha \in \mathcal{I}_{j}} \widetilde{\mathsf{E}}_{N} \left[ \left| \widetilde{\mathsf{E}}_{N} \left[ X_{\alpha} | \mathcal{H}_{2} \right] - p_{\alpha} \right| \right].$$

We define  $U_{\alpha} := V_N \setminus (\mathcal{I} \setminus B_{\alpha})$ . By the Markov property of the DGFF

$$\widetilde{\mathsf{E}}_N\left[X_\alpha \mid \mathcal{H}_2\right] = \widetilde{\mathsf{P}}_{U_\alpha}(\xi_\alpha + h_\alpha \in u_N(R_i)) \quad \widetilde{\mathsf{P}}_N - a. \ s. \tag{4.3}$$

for  $(\xi_{\alpha})_{\alpha \in \mathbb{Z}^d}$  a DGFF with law  $\widetilde{\mathsf{P}}_{U_{\alpha}}$  and  $(h_{\alpha})_{\alpha \in \mathbb{Z}^d}$  is independent of  $\xi$ . From [9, Equation (28)] we can see that, for any  $\alpha \in V_N^{\delta}$  and N large enough such that  $B\left(\alpha,\,(\log N)^{2(1+\epsilon)}\right) \subsetneq V_N$ ,

$$\begin{aligned} \operatorname{Var}\left[h_{\alpha}\right] &= \sum_{\beta \in \mathcal{I} \backslash B_{\alpha}} \mathbb{P}_{\alpha}\left(H_{\mathcal{I} \backslash B_{\alpha}} < +\infty, \, S_{H_{\mathcal{I} \backslash B_{\alpha}}} = \beta\right) g_{N}(\alpha, \, \beta) \\ &\leq \sup_{\beta \in \mathcal{I} \backslash B_{\alpha}} g_{N}(\alpha, \, \beta) \leq \frac{c}{(\log N)^{2(1+\epsilon)(d-2)}}. \end{aligned}$$

This yields

$$\sum_{j=1}^{k} \sum_{\alpha \in \mathcal{I}_{j}} \widetilde{\mathsf{E}}_{N} \left[ \left| \widetilde{\mathsf{P}}_{U_{\alpha}}(\xi_{\alpha} + h_{\alpha}) > u_{N}(R_{j}) \right) - p_{\alpha} \right| \mathbb{1}_{\left\{ |h_{\alpha}| > (u_{N}(z))^{-1 - \epsilon} \right\}} \right] \to 0. \tag{4.4}$$

It then suffices to show

$$\sum_{j=1}^{k} \sum_{\alpha \in \mathcal{I}_{j}} \widetilde{\mathsf{E}}_{N} \left[ \left| \widetilde{\mathsf{P}}_{U_{\alpha}}(\xi_{\alpha} + h_{\alpha}) > u_{N}(R_{j}) \right) - p_{\alpha} \right| \mathbb{1}_{\left\{ |h_{\alpha}| \leq (u_{N}(z))^{-1 - \epsilon} \right\}} \right] \to 0. \tag{4.5}$$

One sees that the breaking up (3.9) can be performed also here replacing  $\varphi_{\alpha}$  and  $\psi_{\alpha}$  (with their laws) with  $\psi_{\alpha}$  and  $\xi_{\alpha}$  (with their laws) respectively, and  $\mu_{\alpha}$  with  $h_{\alpha}$ . Accordingly, it is enough to show that

$$\sum_{\alpha \in \mathcal{I}} \widetilde{\mathsf{E}}_{N} \left[ \left| \widetilde{\mathsf{P}}_{U_{\alpha}}(\xi_{\alpha} + h_{\alpha} > u_{N}(w)) - \widetilde{\mathsf{P}}_{N}(\psi_{\alpha} > u_{N}(w)) \right| \, \mathbb{1}_{\left\{ |h_{\alpha}| \leq (u_{N}(z))^{-1-\epsilon} \right\}} \right] \to 0 \qquad \textbf{(4.6)}$$

for all  $w \in \mathbb{R}$ . To this aim, we choose for any  $w \in \mathbb{R}$  the event

$$\mathcal{Q}' := \left\{ \widetilde{\mathsf{P}}_N(\psi_\alpha > u_N(w)) > \widetilde{\mathsf{P}}_{U_\alpha}(\xi_\alpha + h_\alpha > u_N(w)) \right\}$$

and we proceed as in (3.11) with the help of Lemma 2.1 to show (4.6). Given this, the convergence  $b_3 \to 0$  is finally ensured. Hence we can conclude that

$$\|\mathcal{L}(W_1, \ldots, W_k) - \mathcal{L}(Z_1, \ldots, Z_k)\|_{TV} \to 0$$

where  $Z_i$  are i. i. d. Poisson of mean  $p_{\alpha}$ . By Mills ratio, as in (4.2) we see that

$$P(Z_j = 0) \to \exp\left(-|A_j \cap [\delta, 1 - \delta]^d | \omega(R_j)\right).$$

From this it follows that the two conditions i) and ii) of Kallenberg's Theorem are satisfied, and thus we obtain the convergence to a Poisson point process with mean measure given in i).

*Proof of Theorem 1.2.*  $\mathcal{M}_p(E)$  is a Polish space with metric  $d_p$ :

$$d_p(\mu, \mu') = \sum_{i>1} \frac{\min\{|\mu(f_i) - \mu'(f_i)|, 1\}}{2^i}, \quad \mu, \mu' \in \mathcal{M}_p(E)$$

for a sequence of functions  $f_i \in \mathcal{C}_K^+(E)$  (cf. [12, Section 3.3]). Therefore we are in the condition to use a converging-together theorem [13, Theorem 3.5], namely to prove that  $\rho_n \stackrel{d}{\to} \eta$  it is enough to show the following:

- (a)  $\rho_n^{\delta} \stackrel{d}{\to} \rho^{\delta}$ , as  $n \to +\infty$ .
- (b)  $\rho^{\delta} \stackrel{d}{\to} \eta$  as  $\delta \to 0$ .
- (c) For every  $\epsilon > 0$ ,

$$\lim_{\delta \to 0} \lim_{n \to +\infty} \widetilde{\mathsf{P}}_{N} \left( d_{p} \left( \rho_{n}, \rho_{n}^{\delta} \right) > \epsilon \right) = 0. \tag{4.7}$$

Note that by Lemma 4.1, (a) is satisfied. For  $f \in \mathcal{C}^+_K(E)$ , the Laplace functional of  $\rho^{\delta}$  is given by (cf. [12, Prop. 3.6])

$$\Psi_\delta(f) := \mathsf{E}\left[\exp\left(-\rho^\delta(f)\right)\right] = \exp\left(-\int_E \left(1 - \mathrm{e}^{-f(t,x)}\right) \mathrm{d}\,t_{|_{[\delta,1-\delta]^d}} \,\mathrm{e}^{-x}\,\mathrm{d}\,x\right).$$

Hence by the dominated convergence theorem we can exchange limit and expectation as  $\delta \to 0$  to obtain that

$$\Psi_{\delta}(f) \to \exp\left(-\int_{E} \left(1 - e^{-f(t,x)}\right) dt e^{-x} dx\right)$$

and the right hand side is the Laplace functional of  $\eta$  at f. This shows (b).

Hence to complete the proof it is enough to show (4.7). Thanks to the definition of the metric  $d_p$  it suffices to prove that for  $f \in \mathcal{C}_K^+(E)$  and for  $\epsilon > 0$ 

$$\limsup_{\delta \to 0} \lim_{n \to +\infty} \widetilde{\mathsf{P}}_N \left( \left| \rho_n(f) - \rho_n^{\delta}(f) \right| > \epsilon \right) = 0.$$

Without loss of generality assume that the support of f is contained in  $[0,1]^d \times [z_0, +\infty)$  for some  $z_0 \in \mathbb{R}$ . Choosing n large enough such that  $u_N(z_0) > 0$  and  $g_N(\alpha) \leq g(0)$ , we

obtain that

$$\begin{aligned} \widetilde{\mathsf{E}}_{N} \left[ \left| \rho_{n}(f) - \rho_{n}^{\delta}(f) \right| \right] &= \widetilde{\mathsf{E}}_{N} \left[ \sum_{\alpha \in V_{N} \setminus V_{N}^{\delta}} f\left(\frac{\alpha}{n}, \frac{\psi_{\alpha} - b_{N}}{a_{N}}\right) \mathbb{1}_{\left\{\frac{\psi_{\alpha} - b_{N}}{a_{N}} > z_{0}\right\}} \right] \\ &\leq \sup_{z \in E} |f(z)| \sum_{\alpha \in V_{N} \setminus V_{N}^{\delta}} \widetilde{\mathsf{P}}_{N} \left( \frac{\psi_{\alpha} - b_{N}}{a_{N}} > z_{0} \right) \overset{(3.2)}{\leq} C \sum_{\alpha \in V_{N} \setminus V_{N}^{\delta}} \frac{e^{-u_{N}(z_{0})^{2}/g(0)}}{\sqrt{2\pi}u_{N}(z_{0})} \sqrt{g(0)} \\ &\leq C' \left( 1 - (1 - 2\delta)^{d} \right) e^{-z_{0}} \end{aligned}$$

as  $n \to +\infty$  for some positive constants C, C'. Now letting  $\delta \to 0$  the result follows and this completes the proof.

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