

# ABOUT SCALING BEHAVIOR OF RANDOM BALLS MODELS

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**Abstract.** We study the limit of a generalized random field generated by uniformly scattered random balls as the mean radius of the balls tends to 0 or infinity. Assuming that the radius distribution has a power law behavior, we prove that the centered field, conveniently renormalized, admits a limit.

**Keywords:** Random field, random set, overlapping spheres, Poisson point process, self-similarity.

## 1 Setting

We propose a first step to a unified frame including and extending both the situations of [4] and [2]. We start with a family of grains  $X_j + B(0, R_j)$  in  $\mathbb{R}^d$  built up from a Poisson point process  $(X_j, R_j)_j$  in  $\mathbb{R}^d \times \mathbb{R}^+$ . Equivalently one can start with a Poisson random measure  $N$  on  $\mathbb{R}^d \times \mathbb{R}^+$  and associate with each random point  $(x, r) \in \mathbb{R}^d \times \mathbb{R}^+$  the random ball of center  $x$  and radius  $r$ . We assume that the intensity measure of  $N$  is given by  $dxF(dr)$  where  $F$  is a  $\sigma$ -finite non-negative measure on  $\mathbb{R}^+$  such that

$$\int_{\mathbb{R}^+} r^d F(dr) < +\infty. \quad (1)$$

### 1.1 Assumptions on $F$

We introduce the asymptotic power law behavior assumption on  $F$ , near 0 or at infinity, we will use in the following. For  $\beta > 0$  with  $\beta \neq d$ ,

$$\mathbf{H}(\beta) : F(dr) = f(r)dr \text{ with } f(r) \sim C_\beta r^{-\beta-1}, \text{ as } r \rightarrow 0^{d-\beta},$$

where by convention  $0^\alpha = 0$  if  $\alpha > 0$  and  $0^\alpha = +\infty$  if  $\alpha < 0$ .

Let us remark that according to (1), it is natural to consider the asymptotic behavior of  $F$  around 0 for  $\beta < d$  and at infinity for  $\beta > d$ .

### 1.2 Random field

We consider random fields defined on a space of measures, in the same spirit as the random functionals of [4] or the generalized random fields of [1] (see therein the links between “generalized random fields” and “punctual random fields”). Let  $\mathcal{M}^1$  denote the

space of signed measures  $\mu$  on  $\mathbb{R}^d$  with finite total variation  $\|\mu\|_1 = |\mu|(\mathbb{R}^d)$ , with  $|\mu|$  the total variation measure of  $\mu$ . Since for all  $\mu \in \mathcal{M}^1$ ,

$$\int_{\mathbb{R}^d \times \mathbb{R}^+} |\mu(B(x, r))| dx F(dr) \leq |B(0, 1)| \times \|\mu\|_1 \times \int_{\mathbb{R}^+} r^d F(dr) < +\infty ,$$

one can introduce the generalized random field  $X$  defined on  $\mathcal{M}^1$  as

$$\mu \mapsto \langle X, \mu \rangle = \int_{\mathbb{R}^d \times \mathbb{R}^+} \mu(B(x, r)) N(dx, dr). \quad (2)$$

We introduce the following definition, which coincides with the usual definition of self-similar punctual random fields.

**Definition 1.1.** *A random field  $X$  defined on  $\mathcal{M}^1$  is said to be self-similar with index  $H$  if*

$$\forall s > 0 , \langle X, \mu_s \rangle \stackrel{fdd}{=} s^H \langle X, \mu \rangle \quad \text{where } \mu_s(A) = \mu(s^{-1}A).$$

## 2 Scaling limit

Let us introduce now the notion of “scaling”, by which we indicate an action: a change of scale acts on the size of the grains. The following procedure is performed in [4] where grains of volume  $v$  are changed by shrinking into grains of volume  $\rho v$  with a small parameter  $\rho$  (“small scaling” behavior). The same is performed in [2] in the homogenization section, but the scaling acts in the opposite way: the radii  $r$  of grains are changed into  $r/\varepsilon$  (which is a “large scaling” behavior). Note also that both scalings are performed in the case of  $\alpha$ -stable measures in [3].

Let us multiply the intensity measure by  $\lambda > 0$  and the radii by  $\rho > 0$ . We denote by  $F_\rho(dr)$  the image measure of  $F(dr)$  by the change of scale  $r \mapsto \rho r$ , and consider the associated random field on  $\mathcal{M}^1$  given by

$$\int_{\mathbb{R}^d \times \mathbb{R}^+} \mu(B(x, r)) N_{\lambda, \rho}(dx, dr) ,$$

where  $N_{\lambda, \rho}(dx, dr)$  is the Poisson random measure with intensity measure  $\lambda dx F_\rho(dr)$  and  $\mu \in \mathcal{M}^1$ . Results are expected concerning the asymptotic behavior of this scaled random balls model when  $\rho \rightarrow 0$  or  $\rho \rightarrow +\infty$ . We choose  $\rho$  as the basic model parameter, consider  $\lambda = \lambda(\rho)$  as a function of  $\rho$ , and define on  $\mathcal{M}^1$  the random field

$$X_\rho(\mu) = \int_{\mathbb{R}^d \times \mathbb{R}^+} \mu(B(x, r)) N_{\lambda(\rho), \rho}(dx, dr) .$$

Then, we are looking for a normalization term  $n(\rho)$  such that the centered field converges in distribution,

$$\frac{X_\rho(\cdot) - \mathbf{E}(X_\rho(\cdot))}{n(\rho)} \xrightarrow{fdd} W(\cdot) \quad (3)$$

and we are interested in the nature of the limit field  $W$ . Let us remark that the random field  $X_\rho$  is linear on each vectorial subspace of  $\mathcal{M}^1$  in the sense that for all  $\mu_1, \dots, \mu_n \in \mathcal{M}^1$  and  $a_1, \dots, a_n \in \mathbb{R}$ , almost surely,

$$X_\rho(a_1\mu_1 + \dots + a_n\mu_n) = a_1X_\rho(\mu_1) + \dots + a_nX_\rho(\mu_n).$$

Hence, the finite-dimensional distributions of the normalized field  $(X_\rho(\cdot) - \mathbf{E}(X_\rho(\cdot))) / n(\rho)$  converge toward  $W$  whenever

$$\mathbb{E} \left( \exp \left( i \frac{X_\rho(\mu) - \mathbf{E}(X_\rho(\mu))}{n(\rho)} \right) \right) \rightarrow \mathbb{E} (\exp (iW(\mu))),$$

for all  $\mu$  in a convenient subspace of  $\mathcal{M}^1$ .

To study the limiting behavior of the normalized field  $(X_\rho(\cdot) - \mathbf{E}(X_\rho(\cdot))) / n(\rho)$  we need to impose some assumptions on the measure  $\mu \in \mathcal{M}^1$ . For  $\alpha > 0$  with  $\alpha \neq d$  let us define the space of measures

$$\mathcal{M}_\alpha = \begin{cases} \{\mu \in \mathcal{M}^1; \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} |z - z'|^{-(\alpha-d)} |\mu|(dz) |\mu|(dz') < +\infty\} & \text{if } \alpha > d \\ \{\mu \in \mathcal{M}^1; \int_{\mathbb{R}^d} |z|^{-(\alpha-d)} |\mu|(dz) < +\infty \text{ and } \int_{\mathbb{R}^d} \mu(dz) = 0\} & \text{if } \alpha < d \end{cases},$$

where  $|\mu|$  is the total variation measure of  $\mu$ . Let us consider the kernel on  $\mathbb{R}^d \times \mathbb{R}^d$ ,

$$K_\alpha(z, z') = \begin{cases} |z - z'|^{-(\alpha-d)}, \text{ for } z \neq z' & \text{if } \alpha > d \\ \frac{1}{2} (|z|^{-(\alpha-d)} + |z'|^{-(\alpha-d)} - |z - z'|^{-(\alpha-d)}) & \text{if } \alpha < d \end{cases}.$$

Then, for  $\alpha \in (d-1, 2d)$  with  $\alpha \neq d$ , for any  $\mu \in \mathcal{M}_\alpha$ , the kernel  $K_\alpha$  is defined and non negative  $|\mu| \times |\mu|$  everywhere, with

$$\int_{\mathbb{R}^d} \int_{\mathbb{R}^d} K_\alpha(z, z') |\mu|(dz) |\mu|(dz') < +\infty.$$

For  $\beta \in (d-1, 2d)$  with  $\beta \neq d$ , we finally define the enlarged spaces

$$\overline{\mathcal{M}}_\beta = \begin{cases} \bigcup_{\alpha \in (\beta, 2d)} \mathcal{M}_\alpha & \text{if } \beta > d \\ \bigcup_{\alpha \in (d-1, \beta)} \mathcal{M}_\alpha & \text{if } \beta < d \end{cases}$$

**Theorem 2.1.** *Let  $F$  be a non-negative measure on  $\mathbb{R}^+$  satisfying  $\mathbf{H}(\beta)$  for  $\beta \in (d-1, 2d)$  with  $\beta \neq d$ . For all positive functions  $\lambda$  such that  $n(\rho) := \sqrt{\lambda(\rho)} \rho^\beta \xrightarrow{\rho \rightarrow 0^{\beta-d}} +\infty$ , the limit*

$$\frac{X_\rho(\mu) - \mathbf{E}(X_\rho(\mu))}{n(\rho)} \xrightarrow[\rho \rightarrow 0^{\beta-d}]{fdd} c_\beta W_\beta(\mu)$$

holds for all  $\mu \in \overline{\mathcal{M}}_\beta$ , in the sense of finite dimensional distributions of the random functionals. Here  $W_\beta$  is the centered Gaussian random linear functional on  $\overline{\mathcal{M}}_\beta$  with

$$\mathbb{E} (W_\beta(\mu) W_\beta(\nu)) = \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} K_\beta(z, z') \mu(dz) \nu(dz'), \quad (4)$$

and  $c_\beta$  is a positive constant depending on  $\beta$ .

**Remark 2.2.** *For  $\beta \in (d-1, 2d)$  with  $\beta \neq d$ , the field  $W_\beta$ , defined on  $\overline{\mathcal{M}}_\beta$ , is  $\frac{d-\beta}{2}$  self-similar with  $\frac{d-\beta}{2} < 0$  for  $\beta \in (d, 2d)$  and  $\frac{d-\beta}{2} \in (0, 1/2)$  for  $\beta \in (d-1, d)$ .*

*Proof.* For  $\mu \in \mathcal{M}^1$  let us define the functions  $\varphi_\rho$  and  $\varphi$  on  $\mathbb{R}^+$  by

$$\varphi_\rho(r) = \int_{\mathbb{R}^d} \Psi \left( \frac{\mu(B(x, r))}{n(\rho)} \right) dx, \quad \text{and } \varphi(r) = -\frac{1}{2} \int_{\mathbb{R}^d} \mu(B(x, r))^2 dx,$$

where

$$\Psi(v) = e^{iv} - 1 - iv. \quad (5)$$

The characteristic function of the normalized field  $(X_\rho(\cdot) - \mathbf{E}(X_\rho(\cdot))) / n(\rho)$  is then given by

$$\mathbb{E} \left( \exp \left( i \frac{X_\rho(\mu) - \mathbf{E}(X_\rho(\mu))}{n(\rho)} \right) \right) = \exp \left( \int_{\mathbb{R}^+} \lambda(\rho) \varphi_\rho(r) F_\rho(dr) \right).$$

According to the power law behavior of the density  $F$  we have the following result, which is inspired by Lemma 1 of [4].

**Lemma 2.3.** *Let  $F$  be a non-negative measure on  $\mathbb{R}^+$  satisfying  $\mathbf{H}(\beta)$  for  $\beta > 0$  with  $\beta \neq d$ . Assume that  $g$  is a continuous function on  $\mathbb{R}^+$  such that for some  $0 < p < \beta < q$ , there exists  $C > 0$  such that*

$$|g(r)| \leq C \min(r^q, r^p).$$

Then

$$\int_{\mathbb{R}^+} g(r) F_\rho(dr) \sim C_\beta \rho^\beta \int_0^\infty g(r) r^{-\beta-1} dr \text{ as } \rho \rightarrow 0^{\beta-d}.$$

We apply Lemma 2.3 with the function  $\varphi$ . Since  $\mu \in \mathcal{M}^1$ , the function  $\varphi$  is continuous on  $\mathbb{R}^+$ . Moreover, the next lemma shows that  $\varphi$  satisfies the required upper bound.

**Lemma 2.4.** *Let  $\alpha \in (d-1, 2d)$  with  $\alpha \neq d$ . If  $\mu \in \mathcal{M}_\alpha$ , then there exists  $c > 0$  such that*

$$\int_{\mathbb{R}^d} \mu(B(x, r))^2 dx \leq c \min(r^d, r^\alpha).$$

Therefore, for  $\beta \in (d-1, 2d)$  with  $\beta \neq d$ , when  $\mu \in \overline{\mathcal{M}_\beta}$

$$\int_{\mathbb{R}^+} \int_{\mathbb{R}^d} \mu(B(x, r))^2 r^{-\beta-1} dx dr < +\infty.$$

*Proof.* Since  $\mu \in \mathcal{M}^1$ ,  $\int_{\mathbb{R}^d} \mu(B(x, r))^2 dx \leq |B(0, 1)| \|\mu\|_1^2 r^d$ . We use Lemma 6 of [4] to conclude for  $\mu \in \mathcal{M}_\alpha$ , with  $\alpha \in (d, 2d)$ . When  $\mu \in \mathcal{M}_\alpha$ , with  $\alpha \in (d-1, d)$ , we conclude using the fact that, since  $\int_{\mathbb{R}^d} \mu(dz) = 0$ , we can write

$$\int_{\mathbb{R}^d} \mu(B(x, r))^2 dx = \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \frac{1}{2} (\phi(z, r) + \phi(z', r) - \phi(z - z', r)) \mu(dz) \mu(dz'),$$

where

$$\phi(z, r) = \int_{\mathbb{R}^d} (\mathbf{1}_{B(x, r)}(z) - \mathbf{1}_{B(x, r)}(0))^2 dx, \quad (6)$$

is such that  $\phi(z, r) \leq 2^d |B(0, 1)| |z|^{d-\alpha} r^\alpha$ , for all  $(z, r) \in \mathbb{R}^d \times \mathbb{R}^+$ .  $\square$

When  $F$  satisfies  $\mathbf{H}(\beta)$  for  $\beta \in (d-1, 2d)$  with  $\beta \neq d$ , let us choose  $\mu \in \overline{\mathcal{M}_\beta}$ . According to Lemma 2.4, there exists  $\alpha > 0$  such that  $|\varphi(r)| \leq c \min(r^d, r^\alpha)$ , with  $\alpha > \beta$  if  $\beta > d$  and  $\alpha < \beta$  if  $\beta < d$ . Therefore  $\varphi$  satisfies the assumptions of Lemma 2.3 and

$$\lim_{\rho \rightarrow 0^{\beta-d}} \int_{\mathbb{R}^+} \varphi(r) \lambda(\rho) n(\rho)^{-2} F_\rho(dr) = C_\beta \int_{\mathbb{R}^+} \varphi(r) r^{-\beta-1} dr,$$

since  $n(\rho) = \sqrt{\lambda(\rho)\rho^\beta}$ .

We define the function  $\Delta_\rho(r) = \lambda(\rho)\varphi_\rho(r) - \lambda(\rho)n(\rho)^{-2}\varphi(r)$  and observe that

$$\Delta_\rho(r) = \lambda(\rho) \int_{\mathbb{R}^d} \left( \Psi \left( \frac{\mu(B(x, r))}{n(\rho)} \right) + \frac{1}{2} \left( \frac{\mu(B(x, r))}{n(\rho)} \right)^2 \right) dx. \quad (7)$$

The following result, inspired by Lemma 2 of [4], will play the role of the Lebesgue's Theorem to ensure convergence of the integrals.

**Lemma 2.5.** *Let  $F$  be a non-negative measure on  $\mathbb{R}^+$  satisfying  $\mathbf{H}(\beta)$  for  $\beta > 0$  with  $\beta \neq d$ . Let  $g_\rho$  be a family of continuous functions on  $\mathbb{R}^+$ . Assume that*

$$\lim_{\rho \rightarrow 0^{\beta-d}} \rho^\beta g_\rho(r) = 0, \quad \text{and} \quad \rho^\beta |g_\rho(r)| \leq C \min(r^p, r^q),$$

for some  $0 < p < \beta < q$  and  $C > 0$ . Then

$$\lim_{\rho \rightarrow 0^{\beta-d}} \int_{\mathbb{R}^+} g_\rho(r) F_\rho(dr) = 0.$$

Let us verify that  $\Delta_\rho$  given by (7) satisfies the assumptions of Lemma 2.5. For  $\mu \in \mathcal{M}^1$ , the function  $\Delta_\rho$  is continuous on  $\mathbb{R}^+$ . Because  $\left| \Psi(v) - \left(-\frac{v^2}{2}\right) \right| \leq \frac{|v|^3}{6}$  and

$$\int_{\mathbb{R}^d} \mu(B(x, r))^3 dx \leq \|\mu\|_1^2 \int_{\mathbb{R}^d} |\mu(B(x, r))| dx \leq C_d \|\mu\|_1^3 r^d,$$

we get

$$|\lambda(\rho)^{-1} n(\rho)^2 \Delta_\rho(r)| \leq \frac{C_d \|\mu\|_1^3}{6} n(\rho)^{-1} r^d.$$

Moreover, since  $|\Psi(v)| \leq \frac{|v|^2}{2}$ , by Lemma 2.4 when  $\mu \in \overline{\mathcal{M}}_\beta$ , there exists  $\alpha \in (\beta, 2d)$  if  $\beta > d$ ,  $\alpha \in (d-1, \beta)$  if  $\beta < d$  such that

$$|\lambda(\rho)^{-1} n(\rho)^2 \Delta_\rho(r)| \leq cr^\alpha.$$

Therefore, when  $\mu \in \overline{\mathcal{M}}_\beta$  for  $\beta \in (d-1, 2d)$ , with  $\beta \neq d$ ,

$$\lim_{\rho \rightarrow 0^{\beta-d}} \int_{\mathbb{R}^+} \Delta_\rho(r) F_\rho(dr) = 0.$$

Since

$$\lim_{\rho \rightarrow 0^{\beta-d}} \rho^{-\beta} \int_{\mathbb{R}^+} \varphi(r) F_\rho(dr) = C_\beta \int_{\mathbb{R}^+} \varphi(r) r^{-\beta-1} dr,$$

we get

$$\lim_{\rho \rightarrow 0^{\beta-d}} \mathbb{E} \left( \exp \left( i \frac{X_\rho(\mu) - \mathbf{E}(X_\rho(\mu))}{n(\rho)} \right) \right) = \mathbb{E}(\exp(iW(\mu))),$$

where  $W(\mu)$  is the centered Gaussian random variable with

$$\mathbb{E}(W(\mu)^2) = C_\beta \int_{\mathbb{R}^+} \int_{\mathbb{R}^d} \mu(B(x, r))^2 r^{-\beta-1} dx dr, \quad (8)$$

which is finite using Lemma 2.4. Let us prove that the covariance of  $W$  satisfies (4). By linearity, it is enough to compute the variance of  $W$ . When  $\beta \in (d, 2d)$ , for  $\mu \times \mu$  almost all  $(z, z') \in \mathbb{R}^d \times \mathbb{R}^d$ , the function  $r \mapsto |B(z, r) \cap B(z', r)|$  is in  $L^1(\mathbb{R}^+, r^{-\beta-1} dr)$ . Therefore we can define the kernel,

$$K(z, z') = \int_{\mathbb{R}^+} |B(z, r) \cap B(z', r)| r^{-\beta-1} dr = C(\beta) K_\beta(z, z').$$

By changing the order of integration in (8), we conclude  $W \stackrel{fdd}{=} cW_\beta$  with  $c = (C_\beta C(\beta))^{1/2}$ . When  $\beta \in (d-1, d)$ , the function  $r \mapsto |B(z, r) \cap B(z', r)|$  is not in  $L^1(\mathbb{R}^+, r^{-\beta-1} dr)$  anymore. However, since  $\mu \in \overline{\mathcal{M}_\beta}$ , we have  $\int_{\mathbb{R}^d} \mu(dz) = 0$ , and we can write

$$\int_{\mathbb{R}^d} \mu(B(x, r))^2 dx = \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \frac{1}{2} (\phi(z, r) + \phi(z', r) - \phi(z - z', r)) \mu(dz) \mu(dz'),$$

where  $\phi(z, r)$ , given by (6) is in  $L^1(\mathbb{R}^+, r^{-\beta-1} dr)$ , for all  $z \in \mathbb{R}^d$ . Therefore we can define the kernel  $K(z, z') = \frac{1}{2} (\kappa(z) + \kappa(z') - \kappa(z - z'))$ , with

$$\kappa(z) = \int_{\mathbb{R}^+} \phi(z, r) r^{-\beta-1} dr = C(\beta) K_\beta(z, z).$$

Then, also in this case  $W \stackrel{fdd}{=} cW_\beta$  with  $c = (C_\beta C(\beta))^{1/2}$ , which concludes the proof.  $\square$

Let us mention that similar arguments allow us to state an intermediate scaling result.

**Theorem 2.6.** *Under the assumptions of Theorem 2.1, when  $\lambda(\rho)\rho^\beta \xrightarrow{\rho \rightarrow 0^{\beta-d}} \sigma_0^{d-\beta}$ , for some  $\sigma_0 > 0$ , the following limit holds in the sense of finite dimensional distributions of the random functionals*

$$X_\rho(\mu) - \mathbf{E}(X_\rho(\mu)) \xrightarrow{fdd} J_\beta(\mu_{\sigma_0}),$$

for all  $\mu \in \overline{\mathcal{M}_\beta}$ . Here  $J_\beta$  is the centered random linear functional on  $\overline{\mathcal{M}_\beta}$  defined as

$$J_\beta(\mu) = \int_{\mathbb{R}^d} \int_{\mathbb{R}^+} \mu(B(x, r)) \widetilde{N}_\beta(dx, dr),$$

where  $\widetilde{N}_\beta$  is a compensated Poisson random measure with intensity  $C_\beta dx r^{-\beta-1} dr$ , and  $\mu_{\sigma_0}$  is defined by  $\mu_{\sigma_0}(A) = \mu(\sigma_0^{-1} A)$ .

We should also obtain what is called the small-grain scaling in [4]. In this particular case, the limit is an independently scattered  $\beta$ -stable random measure on  $\mathbb{R}^d$  with Lebesgue measure and unit skewness.

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