

Computation of the Perimeter of Measurable Sets via their Covariogram. Applications to Random Sets *

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Abstract

The covariogram of a measurable set $A \subset \mathbb{R}^d$ is the function g_A which to each $y \in \mathbb{R}^d$ associates the Lebesgue measure of $A \cap (y + A)$. This paper proves two formulas. The first equates the directional derivatives at the origin of g_A to the directional variations of A . The second equates the average directional derivative at the origin of g_A to the perimeter of A . These formulas, previously known with restrictions, are proved for any measurable set. As a by-product, it is proved that the covariogram of a set A is Lipschitz if and only if A has finite perimeter, the Lipschitz constant being half the maximal directional variation. The two formulas have counterparts for mean covariogram of random sets. They also permit to compute the expected perimeter per unit volume of any stationary random closed set. As an illustration, the expected perimeter per unit volume of stationary Boolean models having any grain distribution is computed.

Keywords: Boolean model, covariogram, directional variation, random closed sets, set covariance, set of finite perimeter, specific variation.

1 Introduction

The object of study of this paper is the *covariogram* g_A of a measurable set $A \subset \mathbb{R}^d$ defined for all $y \in \mathbb{R}^d$ by $g_A(y) = \mathcal{L}^d(A \cap (y + A))$, where \mathcal{L}^d denotes the Lebesgue measure. Note that some authors prefer the terms *set covariance* or *covariance function* [5, 6, 22].

Given the covariogram g_A of an unknown set A , a general inverse problem is to determine the geometric information on A that g_A contains. As an important example, Averkov and Bianchi have recently established Matheron's conjecture: up to a translation and a reflection, convex bodies of \mathbb{R}^2 , that is compact convex sets with non-empty interior, are fully determined by their covariogram (see [3])

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and the references within). Contrary to the above mentioned result, this paper focuses on geometric information which is shown to be contained in the covariogram of any measurable set: the perimeter.

As our main results will demonstrate, the perimeter which can be computed from the covariogram is the one from the theory of functions of bounded variation [2]. In this framework, the perimeter of a set A is defined by

$$\text{Per}(A) = \sup \left\{ \int_{\mathbb{R}^d} \mathbb{1}_A(x) \operatorname{div} \varphi(x) dx : \varphi \in \mathcal{C}_c^1(\mathbb{R}^d, \mathbb{R}^d), \|\varphi\|_\infty \leq 1 \right\},$$

and the directional variation in the direction $u \in S^{d-1}$ of A is [2, Section 3.11]

$$V_u(A) = \sup \left\{ \int_{\mathbb{R}^d} \mathbb{1}_A(x) \langle \nabla \varphi(x), u \rangle dx : \varphi \in \mathcal{C}_c^1(\mathbb{R}^d, \mathbb{R}), \|\varphi\|_\infty \leq 1 \right\},$$

where $\mathcal{C}_c^1(\mathbb{R}^d, \cdot)$ denotes the set of continuously differentiable functions with compact support. The non-specialist reader may ask how the perimeter $\text{Per}(A)$ is related to the $(d-1)$ -Hausdorff measure \mathcal{H}^{d-1} of the topological boundary ∂A , which one might consider to be the intuitive notion of surface area. Let us recall that if A is a compact set with Lipschitz boundary (e.g. A is a convex body), then $\text{Per}(A) = \mathcal{H}^{d-1}(\partial A)$, whereas in the general case we only have $\text{Per}(A) \leq \mathcal{H}^{d-1}(\partial A)$ [2, Proposition 3.62]. More precisely, if one defines the essential boundary $\partial_e A$ of A as the set of points of \mathbb{R}^d which are neither Lebesgue density points of A nor of the complementary of A , then $\partial_e A \subset \partial A$ and $\text{Per}(A) = \mathcal{H}^{d-1}(\partial_e A) \leq \mathcal{H}^{d-1}(\partial A)$ [2, Eq. (3.62)]. As shown in [7], the same conclusion holds for directional variations: defining the projection measure μ_u in the direction $u \in S^{d-1}$ by

$$\mu_u(B) = \int_{(\mathbb{R}u)^\perp} \mathcal{H}^0(B \cap (x + \mathbb{R}u)) \mathcal{H}^{d-1}(dx), \quad B \subset \mathbb{R}^d \text{ measurable},$$

one has $V_u(A) = \mu_u(\partial_e A) \leq \mu_u(\partial A)$. In particular, if A is a convex body then $V_u(A) = 2\mathcal{H}^{d-1}(p_u(A))$, where p_u denotes the orthogonal projection with direction u .

Results We prove that for every measurable set A of finite Lebesgue measure,

$$\lim_{r \rightarrow 0} \frac{g_A(0) - g_A(ru)}{|r|} = \frac{1}{2} V_u(A), \quad u \in S^{d-1}. \quad (1)$$

In addition, noting $(g_A^u)'(0+) := \lim_{r \rightarrow 0+} \frac{g_A(ru) - g_A(0)}{r}$ the right directional derivatives at the origin of the covariogram, it is shown that

$$\text{Per}(A) = -\frac{1}{\omega_{d-1}} \int_{S^{d-1}} (g_A^u)'(0+) \mathcal{H}^{d-1}(du), \quad (2)$$

where ω_{d-1} denotes the Lebesgue measure of the unit ball in \mathbb{R}^{d-1} . Hence, for any measurable set A , the perimeter $\text{Per}(A)$ can be computed from the directional derivatives at the origin of the covariogram g_A . As a by-product, it is also shown that a measurable set A has finite perimeter if and only if its covariogram g_A is Lipschitz, and in this case the Lipschitz constant is given by

$$\text{Lip}(g_A) = \frac{1}{2} \sup_{u \in S^{d-1}} V_u(A).$$

Previous work Formula (1) has already been proved for certain classes of sets. It was well-known by the mathematical morphology school [16, 9, 19, 20] that the right directional derivative at the origin of the covariogram g_A of a convex body equals minus the surface area of the orthogonal projection of the set A . The convexity assumption was relaxed in [22] where Rataj extends the result to compact sets in \mathcal{U}_{PR} satisfying a condition of full-dimensionality, \mathcal{U}_{PR} being the family of locally finite unions of sets with positive reach such that all their finite intersections also have positive reach¹. In this more general framework, the surface area of the orthogonal projection is replaced by the total projection $TP_u(A)$ of A , the directional analogue of the $(d-1)$ -total curvature $\Phi_{d-1}(A)$ of A [8]. Formula (1) thus implies that if A is a full-dimensional compact \mathcal{U}_{PR} -set then $V_u(A) = 2TP_u(A)$. This identity is the directional equivalent of a recent result due to Ambrosio, Colesanti and Villa [1]: a full-dimensional compact set with positive reach A satisfies $\text{Per}(A) = 2\Phi_{d-1}(A)$ [1, Theorem 9] (one could directly prove that $V_u(A) = 2TP_u(A)$ by using the techniques developed in [1] and [22]). Since Formula (1) is valid for any measurable set A such that $\mathcal{L}^d(A) < +\infty$, one can argue that the directional variation is the relevant general concept when it comes to the derivative at the origin of the covariogram.

Formula (2) has been widely stated in the mathematical morphology literature [9, 19, 26, 15], under (more or less explicit) regularity assumptions on the set A . We rigorously show that it is valid for any measurable set A having finite Lebesgue measure, provided the perimeter $\text{Per}(A)$ is understood as the variation of A .

The Lipschitz continuity of the covariogram seems to have received less attention in the literature. It is stated in [20] that the covariogram of a compact convex set is Lipschitz and the upper bound of the Lipschitz constant given by Matheron is twice the actual value of this constant.

Applications The covariogram is of particular importance in stochastic geometry when dealing with random closed sets (RACS) [19, 27, 21, 25]. In this context, one defines the mean covariogram of a RACS X as the function $\gamma_X(y) = \mathbb{E}(\mathcal{L}^d(X \cap (y + X)))$. The mean covariogram of a RACS X is related to the probability that two given points belong to X according to the following relation

$$\gamma_X(y) = \int_{\mathbb{R}^d} \mathbb{P}(x \in X \text{ and } x + y \in X) dx.$$

As a consequence the mean covariogram is systematically involved in second order statistics of classic germ-grain models, such as the Boolean model [19, 27, 25], the shot noise model [24, 10], or the dead leaves model [18, 14, 15, 4].

All the established properties of covariograms of deterministic sets extend to the case of mean covariograms of random closed sets. In particular, the stochastic equivalent of Formulas (1) and (2) show that the expectations of the variations of a RACS X are proportional to the directional derivatives of its mean covariogram γ_X (see property 8 of Proposition 16).

If X is any stationary RACS, then its mean covariogram only takes values in $\{0, +\infty\}$ and thus is always degenerate. Nevertheless Formulas (1) and (2) also permit to study the mean variation of stationary RACS. Define the *specific*

¹We refer to [8] and [23] for definitions and results regarding sets with positive reach and \mathcal{U}_{PR} -sets respectively.

directional variation $\theta_{V_u}(X)$ of X as the mean amount of directional variation of X per unit volume (see Section 5.2 for a detailed definition). For any stationary RACS X , it is shown using Eq. (1) that

$$\theta_{V_u}(X) = 2 \lim_{r \rightarrow 0} \frac{1}{|r|} \mathbb{P}(ru \in X, 0 \notin X).$$

Again, integrating over all directions u , one obtains an expression of the *specific variation* $\theta_V(X)$ of X (i.e. the mean amount of variation of X per unit volume)

$$\theta_V(X) = \frac{1}{\omega_{d-1}} \int_{S^{d-1}} \lim_{r \rightarrow 0} \frac{1}{|r|} \mathbb{P}(ru \in X, 0 \notin X) \mathcal{H}^{d-1}(du).$$

As for Eq. (2), the above formula has been stated in the early works of Matheron [17, p. 30] [15, p. 26], but assumptions on the regularity of the RACS were not clearly formulated. It should be emphasized that the specific variation is well-defined for any stationary RACS, and that it can be easily computed as soon as one knows the probabilities $\mathbb{P}(ru \in X, 0 \notin X)$. As an illustration, the specific directional variations and the specific variation of stationary Boolean models are computed in this paper. The obtained expressions generalize known statistics of Boolean models with convex grains [25]. Because it is well-defined for any stationary RACS and easily computable, we claim that, when dealing with non negligible RACS, the specific variation is an interesting alternative to other extension of the usual specific surface area that derives from Steiner's formula [25].

Contents In Section 2 the covariogram g_A of a Lebesgue measurable set A is defined and several properties of g_A are recalled and established. In particular it is shown that as soon as A is non negligible its covariogram g_A is a strictly positive-definite function. Section 3 gathers several known results from the theory of functions of bounded directional variation. In Section 4, the main results relating both the derivative at the origin and the Lipschitz continuity of the covariogram of a set to its directional variations and its perimeter are stated. Finally, applications of these results to the theory of random closed sets are discussed and illustrated in Section 5.

2 Covariogram of a measurable set

Definition 1 (Covariogram of a measurable set). *Let $A \subset \mathbb{R}^d$ be a \mathcal{L}^d -measurable set of finite Lebesgue measure. The covariogram of A is the function $g_A : \mathbb{R}^d \rightarrow [0, +\infty)$ defined for all $y \in \mathbb{R}^d$ by*

$$g_A(y) = \mathcal{L}^d(A \cap (y + A)) = \int_{\mathbb{R}^d} \mathbb{1}_A(x) \mathbb{1}_A(x - y) dx.$$

As initially noted by Matheron [16], the covariogram of A can be expressed as the convolution of the indicator functions of A and its symmetric $\tilde{A} = \{-x \mid x \in A\}$:

$$g_A = \mathbb{1}_A * \mathbb{1}_{\tilde{A}}.$$

As illustrated in the following proposition, this point of view is useful to establish some analytic properties of g_A .

Proposition 2. *Let $A \subset \mathbb{R}^d$ be a \mathcal{L}^d -measurable set of finite Lebesgue measure and g_A be its covariogram. Then*

1. For all $y \in \mathbb{R}^d$, $0 \leq g_A(y) \leq g_A(0) = \mathcal{L}^d(A)$.
2. g_A is even: for all $y \in \mathbb{R}^d$, $g_A(-y) = g_A(y)$.
3. $\int_{\mathbb{R}^d} g_A(y) dy = \mathcal{L}^d(A)^2$.
4. g_A is uniformly continuous over \mathbb{R}^d and $\lim_{|y| \rightarrow +\infty} g_A(y) = 0$.

Proof. The proofs of the three first properties are straightforward. Since $\mathbb{1}_A$ and $\mathbb{1}_{\bar{A}}$ are in $L^2(\mathbb{R}^d)$, the fourth property is obtained in applying the L^p - $L^{p'}$ -convolution theorem to $g_A = \mathbb{1}_A * \mathbb{1}_{\bar{A}}$ (see [11, Proposition 3.2 p. 171] for example). \square

It is well-known that the covariogram is a positive-definite function [16, p. 22], [15, p. 23]. The next proposition improves slightly this result. In particular, it shows that for all $x \neq 0$, $g_A(x) < g_A(0)$.

Proposition 3 (Strict positive-definiteness of the covariogram). *Let A be a \mathcal{L}^d -measurable set such that $0 < \mathcal{L}^d(A) < +\infty$. Then its covariogram g_A is a strictly positive-definite function, that is, for all integers $p \geq 1$, for all p -tuples (y_1, \dots, y_p) of distinct vectors of \mathbb{R}^d , and for all $(w_1, \dots, w_p) \in \mathbb{R}^p \setminus \{0\}$ we have*

$$\sum_{j,k=1}^p w_j w_k g_A(y_k - y_j) > 0.$$

The proof of Proposition 3 makes use of the following lemma.

Lemma 4 (The translations of an integrable function are linearly independent). *Let f be a non null function of $L^1(\mathbb{R}^d)$ and let y_1, \dots, y_p be p distinct vectors of \mathbb{R}^d . Then the functions $x \mapsto f(x + y_j)$, $j = 1, \dots, p$, are linearly independent in $L^1(\mathbb{R}^d)$.*

Proof. Let $(w_1, \dots, w_p) \in \mathbb{R}^p$ be such that $\sum_{j=1}^p w_j f(x + y_j) = 0$ for a.e. $x \in \mathbb{R}^d$.

Applying the Fourier transform we have

$$\left(\sum_{j=1}^p w_j e^{i\langle \xi, y_j \rangle} \right) \hat{f}(\xi) = 0 \text{ for all } \xi \in \mathbb{R}^d.$$

Since f is non null and integrable, \hat{f} is non null and continuous. Hence there exists an open ball $B = B(\xi_0, r)$ of center $\xi_0 \in \mathbb{R}^d$ and radius $r > 0$ such that for all $\xi \in B$, $\hat{f}(\xi) \neq 0$, and thus for all $\xi \in B$, $S(\xi) := \sum_{j=1}^p w_j e^{i\langle \xi, y_j \rangle} = 0$.

One easily shows that the sum $S(\xi)$ is null for all $\xi \in \mathbb{R}^d$ in considering the one-dimensional restriction of S on the line containing ξ and ξ_0 : by the identity theorem, this one-dimensional function is null since it is analytic and null over

an open interval. Applying the inverse generalized Fourier transform to $S = 0$ shows that $\sum_{j=1}^p w_j \delta_{y_j} = 0$. This implies that $w_1 = \dots = w_p = 0$, since by hypothesis the vectors y_j are distinct. \square

Proof of Proposition 3. By Lemma 4, the function $x \mapsto \sum_{j=1}^p w_j \mathbb{1}_A(x + y_j)$ is not a.e. equal to 0. Hence

$$\begin{aligned} \sum_{j,k=1}^p w_j w_k g_A(y_k - y_j) &= \sum_{j,k=1}^p w_j w_k \int_{\mathbb{R}^d} \mathbb{1}_A(x) \mathbb{1}_A(x - y_k + y_j) dx \\ &= \sum_{j,k=1}^p w_j w_k \int_{\mathbb{R}^d} \mathbb{1}_A(x + y_k) \mathbb{1}_A(x + y_j) dx \\ &= \int_{\mathbb{R}^d} \left(\sum_{j=1}^p w_j \mathbb{1}_A(x + y_j) \right)^2 dx > 0. \end{aligned}$$

\square

Proposition 5. *Let $A \subset \mathbb{R}^d$ be a \mathcal{L}^d -measurable set of finite Lebesgue measure and let g_A be its covariogram. Then for all $y, z \in \mathbb{R}^d$*

$$|g_A(y) - g_A(z)| \leq g_A(0) - g_A(y - z).$$

Proof. First let us show that for all measurable sets A_1, A_2 , and A_3

$$\mathcal{L}^d(A_1 \cap A_2) - \mathcal{L}^d(A_1 \cap A_3) \leq \mathcal{L}^d(A_2 \setminus A_3) = \mathcal{L}^d(A_2) - \mathcal{L}^d(A_3 \cap A_2). \quad (3)$$

We have

$$\begin{aligned} \mathcal{L}^d(A_1 \cap A_2) - \mathcal{L}^d(A_1 \cap A_3) &\leq \mathcal{L}^d(A_1 \cap A_2) - \mathcal{L}^d(A_1 \cap A_2 \cap A_3) \\ &= \mathcal{L}^d((A_1 \cap A_2) \setminus (A_1 \cap A_2 \cap A_3)). \end{aligned}$$

Now using that $(A_1 \cap A_2) \setminus (A_1 \cap A_2 \cap A_3)$ is included in the set $A_2 \setminus A_3$, Eq. (3) is proved. Applying Eq. (3) to the sets $A_1 = A$, $A_2 = y + A$ and $A_3 = z + A$ we get

$$\begin{aligned} g_A(y) - g_A(z) &= \mathcal{L}^d(A \cap (y + A)) - \mathcal{L}^d(A \cap (z + A)) \\ &\leq \mathcal{L}^d(y + A) - \mathcal{L}^d((y + A) \cap (z + A)) \\ &= \mathcal{L}^d(A) - \mathcal{L}^d(A \cap ((z - y) + A)) \\ &= g_A(0) - g_A(z - y). \end{aligned}$$

\square

Remark.

- The weaker inequality

$$|g_A(y) - g_A(z)| \leq 2(g_A(0) - g_A(y - z))$$

was established by Matheron [20, p. 1].

- The inequality of Proposition 5 shows that the Lipschitz continuity of the covariogram only depends on the behavior of the function in 0.

3 Facts from the theory of functions of bounded directional variation

This section gathers necessary results from the theory of functions of bounded variation. For a general treatment of the subject we refer to the textbook of Ambrosio, Fusco and Pallara [2]. When the enunciated properties of functions of bounded variation are not found in [2], full proofs are given for the convenience of the reader. Let us add that these proofs can be skipped without impeding on the understanding of the main results of the paper (see Section 4).

For any open subset $U \subset \mathbb{R}^d$, $\mathcal{B}(U)$ denotes the set of Borel subsets of U , and we write $V \subset\subset U$ if $V \subset U$ is open and relatively compact in U .

Definition 6 (Functions of bounded variation). *Let U be an open set of \mathbb{R}^d . We say that $f \in L^1_{\text{loc}}(U)$ is a function of locally bounded variation in U if the distributional derivative of f is representable by a \mathbb{R}^d -valued Radon measure, i.e. if there exists a \mathbb{R}^d -valued Radon measure, noted $Df = (D_1f, \dots, D_df)$, such that for all $\varphi = (\varphi_1, \dots, \varphi_d) \in \mathcal{C}^\infty_c(U, \mathbb{R}^d)$*

$$\int_U f(x) \operatorname{div} \varphi(x) dx = - \sum_{i=1}^d \int_U \varphi_i(x) D_i f(dx). \quad (4)$$

The vector space of all functions of locally bounded variation in U is denoted by $BV_{\text{loc}}(U)$. The functions $f \in BV_{\text{loc}}(U)$ such that $f \in L^1(U)$ and $|Df|(U) < +\infty$ are said to be functions of bounded variation in U and the corresponding function space is denoted by $BV(U)$.

In what follows, S^{d-1} denotes the unit Euclidean sphere in \mathbb{R}^d . If $\varphi \in \mathcal{C}^1(U, \mathbb{R})$ and $u \in S^{d-1}$, we write

$$\frac{\partial \varphi}{\partial u}(x) = \langle \nabla \varphi(x), u \rangle, \quad x \in \mathbb{R}^d,$$

for the directional derivative of φ in the direction u .

Definition 7 (Functions of bounded directional variation). *Let U be an open set of \mathbb{R}^d and let $u \in S^{d-1}$. $f \in L^1_{\text{loc}}(U)$ is a function of locally bounded directional variation in the direction u in U if the directional distributional derivative of f in the direction u is representable by a signed Radon measure, i.e. if there exists a signed Radon measure, noted $D_u f$, such that for all $\varphi \in \mathcal{C}^\infty_c(U, \mathbb{R})$*

$$\int_U f(x) \frac{\partial \varphi}{\partial u}(x) dx = - \int_U \varphi(x) D_u f(dx).$$

The corresponding space is denoted by $BV_{u, \text{loc}}(U)$. The functions $f \in BV_{u, \text{loc}}(U)$ such that $f \in L^1(U)$ and $|D_u f|(U) < +\infty$ are said to be functions of bounded directional variation in the direction u in U and the corresponding function space is denoted by $BV_u(U)$.

The variation in U of a function $f \in L^1_{\text{loc}}(U)$ is defined by

$$V(f, U) = \sup \left\{ \int_U f(x) \operatorname{div} \varphi(x) dx : \varphi \in \mathcal{C}^1_c(U, \mathbb{R}^d), \|\varphi\|_\infty \leq 1 \right\}.$$

A fundamental result of the theory of function of bounded variation states that the variation $V(f, U)$ of $f \in L^1(U)$ is finite if and only if $f \in BV(U)$ [2]. More precisely,

$$V(f, U) = \begin{cases} |Df|(U) & \text{if } f \in BV(U), \\ +\infty & \text{otherwise.} \end{cases}$$

Similarly, for all $f \in L^1_{\text{loc}}(U)$ one defines the *directional variation* in the direction u of f by

$$V_u(f, U) = \sup \left\{ \int_U f(x) \frac{\partial \varphi}{\partial u}(x) dx : \varphi \in \mathcal{C}_c^1(U, \mathbb{R}), \|\varphi\|_\infty \leq 1 \right\},$$

and

$$V_u(f, U) = \begin{cases} |D_u f|(U) & \text{if } f \in BV_u(U), \\ +\infty & \text{otherwise.} \end{cases}$$

If $A \subset \mathbb{R}^d$ is a \mathcal{L}^d -measurable set, by definition the *perimeter* of A in U is the variation of the indicator function $\mathbb{1}_A$ in U and one notes $\text{Per}(A, U) := V(\mathbb{1}_A, U)$. Besides, one writes $V_u(A, U) := V_u(\mathbb{1}_A, U)$ for the *directional variation* of A in the direction u in U . In the case where $U = \mathbb{R}^d$, one simply writes $V(f) = V(f, \mathbb{R}^d)$ and $V_u(f) = V_u(f, \mathbb{R}^d)$, and similarly for the variations of a set.

As shown by the next proposition, given all the directional variations $V_u(f, U)$, $u \in S^{d-1}$, one can compute the variation $V(f, U)$.

Proposition 8 (Variation and directional variations). *Let U be an open set of \mathbb{R}^d and let $f \in L^1(U)$. Then, the three following assertions are equivalent:*

- (i) $f \in BV(U)$.
- (ii) $f \in BV_u(U)$ for all $u \in S^{d-1}$.
- (iii) For all vectors e_i of the canonical basis, $f \in BV_{e_i}(U)$.

In addition,

$$\frac{1}{d} V(f, U) \leq \frac{1}{d} \sum_{i=1}^d V_{e_i}(f, U) \leq \sup_{u \in S^{d-1}} V_u(f, U) \leq V(f, U),$$

and

$$V(f, U) = \frac{1}{2\omega_{d-1}} \int_{S^{d-1}} V_u(f, U) \mathcal{H}^{d-1}(du), \quad (5)$$

where ω_{d-1} denotes the Lebesgue measure of the unit ball in \mathbb{R}^{d-1} .

The results of this proposition are mostly from [7]. A proof is reproduced below for the convenience of the reader. First one needs the following lemma.

Lemma 9. *Let $f \in L^1(U)$ and let $u_1, \dots, u_k \in S^{d-1}$ such that $f \in BV_{u_j}(U)$ for all $j = 1, \dots, k$. Then for all $u = \sum_j \lambda_j u_j \in S^{d-1} \cap \text{span}\{u_1, \dots, u_k\}$, $f \in BV_u(U)$.*

Proof. With the above notation one easily checks that $D_u f := \sum_j \lambda_j D_{u_j} f$ is a signed Radon measure which represents the directional derivative of f in the direction u . Besides, by the triangle inequality $|D_u f|(U) \leq \sum_j^k |\lambda_j| |D_{u_j} f|(U) < +\infty$, and thus $f \in BV_u(U)$. \square

Proof of Proposition 8. Clearly, Assertion (ii) implies Assertion (iii). Let us show that (i) implies (ii). Let $f \in BV(U)$, let $Df = (D_1f, \dots, D_df)$ be the Radon measure representing its distributional derivative, and let $u \in S^{d-1}$. Then $\langle Df, u \rangle := \sum_{i=1}^d u_i D_i f$ is a signed Radon measure which represents the directional derivative of f in the direction u , and by the Cauchy-Schwarz inequality

$$V_u(f, U) = |\langle Df, u \rangle|(U) \leq \sum_{i=1}^d |u_i| |D_i f|(U) \leq |u| |Df|(U) = |Df|(U) = V(f, U) < +\infty.$$

Hence $f \in BV_u(U)$ and $V_u(f, U) \leq V(f, U)$. Let us now show that (iii) implies (i). For all vectors e_i of the canonical basis, $f \in BV_{e_i}(\mathbb{R}^d)$ and there exists a signed Radon measure $D_{e_i} f$ which represents the distributional partial derivatives of f . But then one easily checks that $(D_{e_1} f, \dots, D_{e_d} f)$ is a \mathbb{R}^d -valued Radon measure which represents the distributional derivative of f . In addition, from the definition of the variation

$$V(f, U) \leq \sum_{i=1}^d V_{e_i}(f, U) < +\infty,$$

and thus $f \in BV(U)$. This concludes the proof of the announced equivalence as well as the proof of the inequalities since

$$V(f, U) \leq \sum_{i=1}^d V_{e_i}(f, U) \leq d \sup_{u \in S^{d-1}} V_u(f, U) \leq dV(f, U).$$

To finish let us show Formula (5). First let us suppose that $f \notin BV(U)$ and let us show that the right-hand side of Formula (5) is equal to $+\infty$. Remark that by Lemma 9 and the equivalence above the set of directions $u \in S^{d-1}$ for which $V_u(f, U) < +\infty$ is contained in a linear subspace of dimension less than $d - 1$ (otherwise f would be in $BV_u(U)$ for all u and consequently f would be in $BV(U)$). Hence for \mathcal{H}^{d-1} -all u in S^{d-1} , $V_u(f, U) = +\infty$, and thus the right-hand side of Eq. (5) equals $+\infty$. Let us now suppose that $f \in BV(U)$. By the polar decomposition theorem [2, Corollary 1.29] there exists a unique $|Df|$ -integrable function $\sigma : U \rightarrow S^{d-1}$ such that $Df = \sigma |Df|$. With this notation, observe that for all $u \in S^{d-1}$ and $A \in \mathcal{B}(U)$,

$$D_u f(A) = \langle Df, u \rangle(A) = \int_A \langle \sigma(x), u \rangle |Df|(dx).$$

Hence, by [2, Proposition 1.23]

$$V_u(f, U) = |D_u f|(U) = \int_U |\langle \sigma(x), u \rangle| |Df|(dx).$$

For all $\nu \in S^{d-1}$ the following well-known identity holds

$$\int_{S^{d-1}} |\langle \nu, u \rangle| \mathcal{H}^{d-1}(du) = 2\omega_{d-1}.$$

Hence by Fubini's theorem

$$\begin{aligned} \int_{S^{d-1}} V_u(f, U) \mathcal{H}^{d-1}(du) &= \int_U \left(\int_{S^{d-1}} |\langle \sigma(x), u \rangle| \mathcal{H}^{d-1}(du) \right) |Df|(dx) \\ &= 2\omega_{d-1} |Df|(U) \\ &= 2\omega_{d-1} V(f, U). \end{aligned}$$

□

The next proposition recalls fundamental properties related to the approximation of functions of bounded directional variation. For simplicity we restrict ourselves to the case $U = \mathbb{R}^d$. See [2, Section 3.11] for the proofs.

Proposition 10 (Directional variation and approximation).

- Variation of smooth functions: *If $f \in \mathcal{C}^1(\mathbb{R}^d) \cap L^1(\mathbb{R}^d)$ then*

$$V_u(f) = \int_{\mathbb{R}^d} \left| \frac{\partial f}{\partial u}(x) \right| dx.$$

- Lower semi-continuity with respect to the L^1 -convergence: *If f_n converges towards f in $L^1(\mathbb{R}^d)$ then $V_u(f) \leq \liminf_{n \rightarrow +\infty} V_u(f_n)$.*
- Approximation by smooth functions: *for every function $f \in BV_u(\mathbb{R}^d)$, there exists a sequence of smooth functions $f_n \in \mathcal{C}^\infty(\mathbb{R}^d) \cap BV_u(\mathbb{R}^d)$ such that f_n converges towards f in $L^1(\mathbb{R}^d)$ and $\lim_{n \rightarrow +\infty} V_u(f_n) = V_u(f)$.*

One practical advantage of directional variations $V_u(f)$ over the non-directional variation $V(f)$ is that it can be computed from the integrals of difference quotients, as the next proposition recalls. Although this is a standard result of BV functions theory², the author is not aware of any standard textbook which enunciates it. Consequently a proof is given for the convenience of the reader.

Proposition 11 (Directional variation and difference quotient). *Let $u \in S^{d-1}$ and let $f \in L^1(\mathbb{R}^d)$ be any integrable function. Then for all $r \neq 0$,*

$$\int_{\mathbb{R}^d} \frac{|f(x+ru) - f(x)|}{|r|} dx \leq V_u(f)$$

and

$$\lim_{r \rightarrow 0} \int_{\mathbb{R}^d} \frac{|f(x+ru) - f(x)|}{|r|} dx = V_u(f).$$

Proof. To prove the inequality we can suppose that $f \in BV_u(\mathbb{R}^d)$. First suppose that $f \in \mathcal{C}^1(\mathbb{R}^d) \cap BV_u(\mathbb{R}^d)$. Then

$$|f(x+ru) - f(x)| = \left| \int_0^1 r \frac{\partial f}{\partial u}(x+tru) dt \right| \leq \int_0^1 |r| \left| \frac{\partial f}{\partial u}(x+tru) \right| dt.$$

²L. Ambrosio, personal communication.

Hence, using Fubini's theorem and the first point of Proposition 10,

$$\int_{\mathbb{R}^d} \frac{|f(x+ru) - f(x)|}{|r|} dx \leq \int_0^1 \left(\int_{\mathbb{R}^d} \left| \frac{\partial f}{\partial u}(x+tru) \right| dx \right) dt = V_u(f).$$

This inequality is shown to be valid for any $f \in BV_u(\mathbb{R}^d)$ by using approximation by smooth functions (see Proposition 10).

Let us now turn to the second part of the statement. Let $f \in L^1(\mathbb{R}^d)$. Using the above inequality it is enough to show that

$$V_u(f) \leq \liminf_{r \rightarrow 0} \int_{\mathbb{R}^d} \frac{|f(x+ru) - f(x)|}{|r|} dx.$$

Let us consider a family of mollifiers $(\rho_\varepsilon)_{\varepsilon > 0}$, that is functions $\rho_\varepsilon(x) := \varepsilon^{-d} \rho(x/\varepsilon)$ where the function ρ is even, non negative, \mathcal{C}^∞ , with support contained in the unit ball, and such that $\int_{\mathbb{R}^d} \rho(x) dx = 1$. Define $f_\varepsilon = f * \rho_\varepsilon$. Then $f_\varepsilon \in \mathcal{C}^\infty(\mathbb{R}^d) \cap L^1(\mathbb{R}^d)$ and f_ε converges towards f in $L^1(\mathbb{R}^d)$ as ε tends to 0. By Fatou's lemma we have

$$V_u(f_\varepsilon) = \int_{\mathbb{R}^d} \left| \frac{\partial f_\varepsilon}{\partial u}(x) \right| dx \leq \liminf_{r \rightarrow 0} \int_{\mathbb{R}^d} \frac{|f_\varepsilon(x+ru) - f_\varepsilon(x)|}{|r|} dx.$$

Since $\|f_\varepsilon(\cdot + ru) - f_\varepsilon(\cdot)\|_1 = \|(f(\cdot + ru) - f(\cdot)) * \rho_\varepsilon\|_1 \leq \|f(\cdot + ru) - f(\cdot)\|_1$ we deduce that for all $\varepsilon > 0$

$$V_u(f_\varepsilon) \leq \liminf_{r \rightarrow 0} \int_{\mathbb{R}^d} \frac{|f(x+ru) - f(x)|}{|r|} dx.$$

Using the lower semi-continuity of the directional variation with respect to the L^1 -convergence we get the result. \square

4 Directional variation, perimeter and covariogram of measurable sets

In this section, the main results of the paper are established (see Theorem 13 and Theorem 14).

Lemma 12 ([20]). *Let A be a \mathcal{L}^d -measurable set having finite Lebesgue measure and let g_A be its covariogram. Then for all $y \in \mathbb{R}^d$*

$$g_A(0) - g_A(y) = \frac{1}{2} \int_{\mathbb{R}^d} |\mathbb{1}_A(x+y) - \mathbb{1}_A(x)| dx.$$

Proof.

$$\int_{\mathbb{R}^d} |\mathbb{1}_A(x+y) - \mathbb{1}_A(x)| dx = \int_{\mathbb{R}^d} (\mathbb{1}_A(x+y) - \mathbb{1}_A(x))^2 dx = 2(g_A(0) - g_A(y)).$$

\square

The identity of Lemma 12, which is due to Matheron [20], is the key point to apply the results from the theory of functions of bounded directional variations enunciated in Section 3. First, one establishes Formula (1) and obtains a characterization of sets of finite directional variation.

Theorem 13 (Directional variation and covariogram of measurable sets). *Let A be a \mathcal{L}^d -measurable set having finite Lebesgue measure, let g_A be its covariogram, and let $u \in S^{d-1}$. The following assertions are equivalent:*

(i) A has finite directional variation $V_u(A)$.

(ii) $\lim_{r \rightarrow 0} \frac{g_A(0) - g_A(ru)}{|r|}$ exists and is finite.

(iii) The one-dimensional restriction of the covariogram $g_A^u : r \mapsto g_A(ru)$ is Lipschitz.

In addition,

$$\text{Lip}(g_A^u) = \lim_{r \rightarrow 0} \frac{g_A(0) - g_A(ru)}{|r|} = \frac{1}{2}V_u(A),$$

the second equality being also valid when $V_u(A) = +\infty$.

Remark. Note that Assertion (ii) of Theorem 13 above can be replaced by “The right directional derivative $g_A^u(0+)$ exists and is finite” since

$$g_A^u(0+) := \lim_{r \rightarrow 0+} \frac{g_A(ru) - g_A(0)}{r} = - \lim_{r \rightarrow 0} \frac{g_A(0) - g_A(ru)}{|r|}.$$

Proof. Since from Lemma 12,

$$\frac{g_A(0) - g_A(ru)}{|r|} = \frac{1}{2} \int_{\mathbb{R}^d} \frac{|\mathbb{1}_A(x + ru) - \mathbb{1}_A(x)|}{|r|} dx,$$

by applying Proposition 11 with $f = \mathbb{1}_A$ one obtains the equivalence of (i) and

(ii) as well as the formula $\lim_{r \rightarrow 0} \frac{g_A(0) - g_A(ru)}{|r|} = \frac{1}{2}V_u(A)$.

Let us show that (i) implies (iii). By Proposition 5, for all r and $s \in \mathbb{R}$

$$|g_A(ru) - g_A(su)| \leq g_A(0) - g_A((r-s)u) = \frac{1}{2} \int_{\mathbb{R}^d} |\mathbb{1}_A(x + (r-s)u) - \mathbb{1}_A(x)| dx.$$

Applying the inequality of Proposition 11 with $f = \mathbb{1}_A$,

$$|g_A(ru) - g_A(su)| \leq \frac{1}{2}|r-s| \int_{\mathbb{R}^d} \frac{|\mathbb{1}_A(x + (r-s)u) - \mathbb{1}_A(x)|}{|r-s|} dx \leq \frac{1}{2}V_u(A)|r-s|.$$

Hence g_A^u is Lipschitz and $\text{Lip}(g_A^u) \leq \frac{1}{2}V_u(A)$.

Let us now show that (iii) implies (i). For all $r \neq 0$ we have

$$\text{Lip}(g_A^u) \geq \frac{g_A(0) - g_A(ru)}{|r|} = \frac{1}{2} \int_{\mathbb{R}^d} \frac{|\mathbb{1}_A(x + ru) - \mathbb{1}_A(x)|}{|r|} dx.$$

By Proposition 11 the right-hand side tends towards $\frac{1}{2}V_u(A)$ as r tends to 0. Hence A has finite directional variation in the direction u and $\text{Lip}(g_A^u) \geq \frac{1}{2}V_u(A)$. All in all we have shown that (i) and (iii) are equivalent and that $\text{Lip}(g_A^u) = \frac{1}{2}V_u(A)$. \square

Considering all the possible directions $u \in S^{d-1}$, the results of the previous theorem lead to Formula (2) (reproduced below as Formula (6)) and a characterization of sets of finite perimeter.

Theorem 14 (Perimeter and covariogram of measurable sets). *Let A be a \mathcal{L}^d -measurable set having finite Lebesgue measure, and let g_A be its covariogram. The following assertions are equivalent:*

(i) A has finite perimeter $\text{Per}(A)$.

(ii) For all $u \in S^{d-1}$, $(g_A^u)'(0+) := \lim_{r \rightarrow 0+} \frac{g_A(ru) - g_A(0)}{r}$ exists and is finite.

(iii) The covariogram g_A is Lipschitz.

In addition the following relations hold:

$$\text{Lip}(g_A) = \frac{1}{2} \sup_{u \in S^{d-1}} V_u(A) \leq \frac{1}{2} \text{Per}(A)$$

and

$$\text{Per}(A) = -\frac{1}{\omega_{d-1}} \int_{S^{d-1}} (g_A^u)'(0+) \mathcal{H}^{d-1}(du), \quad (6)$$

this last formula being also valid when $\text{Per}(A) = +\infty$.

Proof. The equivalence of (i) and (ii) as well as the integral geometric formula of Eq. (6) derive from Proposition 8 and the identity

$$(g_A^u)'(0+) = \lim_{r \rightarrow 0+} \frac{g_A(ru) - g_A(0)}{r} = -\frac{1}{2} V_u(A).$$

Let us now show that (i) implies (iii). Let $y, z \in \mathbb{R}^d$. Denote by u the direction of S^{d-1} such that $y - z = |y - z|u$. By Proposition 5 and Theorem 13

$$|g_A(y) - g_A(z)| \leq g_A(0) - g_A(y - z) \leq \frac{1}{2} V_u(A) |y - z| \leq \left(\frac{1}{2} \sup_{u \in S^{d-1}} V_u(A) \right) |y - z|.$$

Hence g_A is Lipschitz and $\text{Lip}(g_A) \leq \frac{1}{2} \sup_u V_u(A)$. As for the converse implication and inequality, for all $u \in S^{d-1}$,

$$\text{Lip}(g_A) \geq \lim_{r \rightarrow 0} \frac{g_A(0) - g_A(ru)}{|r|} = \frac{1}{2} V_u(A).$$

Hence for all $u \in S^{d-1}$, $V_u(A) < +\infty$ and $\text{Lip}(g_A) \geq \frac{1}{2} \sup_u V_u(A)$. This concludes the proof. \square

One natural question is whether Formula (6) extends to the case of functions, that is if one can recover the variation $V(f)$ of a function f from the directional derivatives of its covariogram $g_f(y) := \int_{\mathbb{R}^d} f(x+y)f(x)dx$. The answer to this question is negative. Indeed, if one considers a smooth function $f \in \mathcal{C}_c^1(\mathbb{R}^d)$, then its covariogram g_f is well-defined and is differentiable in 0. But since g_f is even, its derivative at the origin equals zero, and thus the variation of f is not equal to the integral of the directional derivatives of the covariogram g_f .

5 Applications to random closed sets

5.1 Mean covariogram of a random closed set

A random closed set (RACS) X is a measurable map from a probability space $(\Omega, \mathcal{A}, \mathbb{P})$ to the space $\mathcal{F}(\mathbb{R}^d)$ of closed subsets of \mathbb{R}^d endowed with the σ -algebra generated by the family $\{\{F \in \mathcal{F}(\mathbb{R}^d), F \cap K = \emptyset\}, K \text{ compact}\}$ [19, 21, 27].

Definition 15 (Mean covariogram of a random closed set). *Let X be a random closed set (RACS) of \mathbb{R}^d having finite mean Lebesgue measure, i.e. $\mathbb{E}(\mathcal{L}^d(X)) < +\infty$. The mean covariogram γ_X of X is the expectation of the covariogram of X with respect to its distribution, that is $\gamma_X : \mathbb{R}^d \rightarrow [0, \infty)$ is the function defined by*

$$\gamma_X(y) = \mathbb{E}(g_X(y)) = \mathbb{E}(\mathcal{L}^d(X \cap (y + X))) = \int_{\mathcal{F}(\mathbb{R}^d)} \mathcal{L}^d(A \cap (y + A)) \mathbb{P}_X(dA).$$

As the next proposition will show, all the results relative to covariograms of deterministic measurable sets can be adapted for mean covariograms of RACS. However before stating these results, we need to introduce the notions of mean perimeter $\mathbb{E}(\text{Per}(X))$ and mean directional variations $\mathbb{E}(V_u(X))$, $u \in S^{d-1}$, of a RACS X .

We say that a jointly measurable random field $f : \Omega \times \mathbb{R}^d \rightarrow \mathbb{R}$ almost surely (a.s.) in $L^1_{\text{loc}}(\mathbb{R}^d)$ has a.s. *locally bounded variation* if there exists a random \mathbb{R}^d -valued Radon measure³ Df which represents the distributional derivative of f , i.e. Eq. (4) holds a.s. If in addition $f \in L^1(\mathbb{R}^d)$ a.s. and $|Df|(\mathbb{R}^d)$ is a.s. finite, then f is said to have a.s. *bounded variation* in \mathbb{R}^d . Similarly, $f : \Omega \times \mathbb{R}^d \rightarrow \mathbb{R}$ a.s. in $L^1_{\text{loc}}(\mathbb{R}^d)$ has a.s. *locally bounded directional variation* in the direction $u \in S^{d-1}$ if there exists a random signed Radon measure $D_u f$ representing the distributional directional derivative of f . If $f \in L^1(\mathbb{R}^d)$ a.s. and $|D_u f|(\mathbb{R}^d) < +\infty$ a.s., then one says that f has *bounded directional variation* in the direction u .

The *mean variation* $\mathbb{E}(V(f))$ and the *mean directional variations* $\mathbb{E}(V_u(f))$, $u \in S^{d-1}$, of a random field f a.s. in $L^1(\mathbb{R}^d)$ are defined by

$$\mathbb{E}(V(f)) = \begin{cases} \mathbb{E}(|Df|(\mathbb{R}^d)) & \text{if } f \text{ has a.s. bounded variation,} \\ +\infty & \text{otherwise,} \end{cases}$$

and

$$\mathbb{E}(V_u(f)) = \begin{cases} \mathbb{E}(|D_u f|(\mathbb{R}^d)) & \text{if } f \text{ has a.s. bounded directional variation,} \\ +\infty & \text{otherwise.} \end{cases}$$

Since any RACS X defines a jointly measurable random field by $(\omega, x) \mapsto \mathbb{1}_{X(\omega)}(x)$ (see [21, p. 59]), the mean perimeter $\mathbb{E}(\text{Per}(X))$ and the mean directional variations $\mathbb{E}(V_u(X))$ of a RACS X are well-defined.

Proposition 16 (Properties of the mean covariogram of a RACS). *Let X be a RACS of \mathbb{R}^d satisfying $\mathbb{E}(\mathcal{L}^d(X)) < +\infty$ and let γ_X be its mean covariogram. Then*

³We refer to [12] for definitions relative to random Radon measures and for a proof of the fact that the variation $|\mu|$ of a random Radon measure μ is a well-defined random positive Radon measure.

1. For all $y \in \mathbb{R}^d$, $0 \leq \gamma_X(y) \leq \gamma_X(0) = \mathbb{E}(\mathcal{L}^d(X))$.
2. γ_X is even.
3. $\gamma_X(y) = \int_{\mathbb{R}^d} \mathbb{P}(x \in X \text{ and } x + y \in X) dx$.
4. $\int_{\mathbb{R}^d} \gamma_X(y) dy = \mathbb{E}(\mathcal{L}^d(X)^2) \in [0, +\infty]$.
5. If $\mathbb{E}(\mathcal{L}^d(X)) > 0$, then γ_X is a strictly positive-definite function.
6. For all $y, z \in \mathbb{R}^d$, $|\gamma_X(y) - \gamma_X(z)| \leq \gamma_X(0) - \gamma_X(y - z)$.
7. γ_X is uniformly continuous over \mathbb{R}^d and $\lim_{|y| \rightarrow +\infty} \gamma_X(y) = 0$.
8. We have

$$\lim_{r \rightarrow 0} \frac{\gamma_X(0) - \gamma_X(ru)}{|r|} = \frac{1}{2} \mathbb{E}(V_u(X)),$$

$$\text{and, noting } (\gamma_X^u)'(0+) = \lim_{r \rightarrow 0+} \frac{\gamma_X(ru) - \gamma_X(0)}{r},$$

$$-\frac{1}{\omega_{d-1}} \int_{S^{d-1}} (\gamma_X^u)'(0+) \mathcal{H}^{d-1}(du) = \mathbb{E}(\text{Per}(X)).$$

The proofs are omitted since they mostly consist in integrating the results of the previous sections with respect to the distribution of the RACS X . When $\mathbb{E}(V_u(X)) < +\infty$ and $\mathbb{E}(\text{Per}(X)) < +\infty$, both formulas of property 8 follows easily from the bounded convergence theorem. Using Fatou's lemma, one shows that these formulas are also valid when $\mathbb{E}(V_u(X)) = +\infty$ and $\mathbb{E}(\text{Per}(X)) = +\infty$.

5.2 Specific variation of a stationary RACS

A RACS X is said to be *stationary* if for all $y \in \mathbb{R}^d$, the translated RACS $y + X$ has the same distribution as X . If a RACS X is stationary, one defines its *variogram* ν_X as the function $\nu_X(y) = \mathbb{P}(y \in X, 0 \notin X)$ (see e.g. [15] for more details on variograms).

If a stationary RACS X has locally bounded variation, then one easily checks that its derivative $D\mathbb{1}_X$, which is by definition a random \mathbb{R}^d -valued Radon measure, is also stationary. Consequently, the variation measure $|D\mathbb{1}_X|$ is a stationary positive Radon measure, and thus there exists a real number $\theta_V(X) \in [0, +\infty]$ such that for all nonempty open sets $U \subset \subset \mathbb{R}^d$,

$$\mathbb{E}(V(X, U)) := \mathbb{E}(|D\mathbb{1}_X|(U)) = \theta_V(X) \mathcal{L}^d(U).$$

We choose to call this constant $\theta_V(X)$ the *specific variation* of X or the *variation intensity* of X (see the discussion below). Similarly, for all $u \in S^{d-1}$ there exists a real $\theta_{V_u}(X) \in [0, +\infty]$ such that $\mathbb{E}(|D_u\mathbb{1}_X|(U)) = \theta_{V_u}(X) \mathcal{L}^d(U)$. $\theta_{V_u}(X)$ is called the *specific directional variation* of X in the direction u (or also the *directional variation intensity*). As before, one extends the definition of the specific variation for stationary RACS X which do not have a.s. locally bounded variation by setting $\theta_V(X) = +\infty$, and similarly for the specific directional

variation $\theta_{V_u}(X)$. In this context the integral-geometric formula of Eq. (5) gives

$$\theta_V(X) = \frac{1}{2\omega_{d-1}} \int_{S^{d-1}} \theta_{V_u}(X) \mathcal{H}^{d-1}(du).$$

Theorem 17 (Specific variations and variogram). *Let X be a stationary RACS, let ν_X be its variogram, and, for all $u \in S^{d-1}$, denote $(\nu_X^u)'(0+) := \lim_{r \rightarrow 0+} \frac{1}{r} \nu_X(ru)$. Then for all $u \in S^{d-1}$ the specific directional variation $\theta_{V_u}(X)$ is given by*

$$\theta_{V_u}(X) = 2(\nu_X^u)'(0+) = 2 \lim_{r \rightarrow 0} \frac{1}{|r|} \mathbb{P}(ru \in X, 0 \notin X).$$

In other words, the specific directional variation is twice the right directional derivative of the variogram at the origin. Integrating over all directions, one obtains the specific variation of X :

$$\theta_V(X) = \frac{1}{\omega_{d-1}} \int_{S^{d-1}} (\nu_X^u)'(0+) \mathcal{H}^{d-1}(du). \quad (7)$$

Before proving this theorem let us discuss the terminology *specific variation* of X for the constant $\theta_V(X)$. Eq. (7) is the formula given in [15, p. 26] and which originates from Matheron [17, p. 30]. In these references, the constant corresponding to the variation intensity $\theta_V(X)$ is called the *specific $(d-1)$ -volume* of X (*specific perimeter* if $d=2$, *specific surface area* if $d=3$). However, in the later works of Matheron [19] as well as on recent reference textbooks [27, 25], the *specific surface measure* refers to the surface measure that derives from Steiner's formula. This measure has different names, depending on its normalization and the degree of generalization: intrinsic volume of index $d-1$ and Minkowski's content of index 1 for convex sets [25], total curvature of index $d-1$ for sets with positive reach and \mathcal{U}_{PR} -sets [8, 23], or also, in a more general setting, outer Minkowski content [1, 28]; see also [13]. Even though the (variational) perimeter of a set and this notion of surface measure agree for convex sets [1], the distinction is important. Indeed their extensions to non convex sets have different behaviors. For example, the outer Minkowski content counts twice the isolated fine parts of a set having a bounded and $(d-1)$ -rectifiable topological boundary, whereas these fine parts have no influence on the perimeter [28, Proposition 4.1] (here "isolated fine parts" denotes the part of the boundary which has Lebesgue density 0). In order to make a clear distinction between the (variational) perimeter and the surface measure from Steiner's formula, the constant $\theta_V(X)$ is named the *specific variation* of X and not its "specific perimeter".

As mentioned in the introduction, one should notice that, contrary to the specific surface area [25], the specific variation $\theta_V(X)$ is well-defined for any stationary RACS. Besides, Theorem 17 shows that the specific directional variations $\theta_{V_u}(X)$ and the specific variation $\theta_V(X)$ are easily computed as soon as one knows the variogram of X . This will be illustrated in the next section where the specific variations of stationary Boolean models are computed.

Let us now turn to the proof of Theorem 17 which uses the following intuitive lemma.

Lemma 18. *Let A be a \mathcal{L}^d -measurable set and B be an open ball. Then for all $u \in S^{d-1}$,*

$$V_u(A, B) \leq V_u(A \cap B, \mathbb{R}^d) \leq V_u(A, B) + V_u(B, \mathbb{R}^d).$$

References for the proof. The first inequality is immediate from the definition of the directional variation on an open set [2]. The second inequality is easily proved using the interpretation of the directional variation as the projection measure of the essential boundary mentioned in the introduction [7]. \square

Proof of Theorem 17. First remark that

$$\begin{aligned} \mathbb{P}(ru \in X, 0 \notin X) &= \mathbb{P}(0 \in X, -ru \notin X) \\ &= \mathbb{P}(0 \in X) - \mathbb{P}(0 \in X \text{ and } -ru \in X) \\ &= \mathbb{P}(0 \in X) - \mathbb{P}(0 \in X \cap (ru + X)). \end{aligned}$$

Let B be any nonempty open ball. Since X is a stationary RACS

$$\mathbb{P}(0 \in X) = \frac{\mathbb{E}(\mathcal{L}^d(X \cap B))}{\mathcal{L}^d(B)}.$$

As $X \cap (ru + X)$ is also a stationary RACS, we have

$$\mathbb{P}(0 \in X \cap (ru + X)) = \frac{\mathbb{E}(\mathcal{L}^d(X \cap (ru + X) \cap B))}{\mathcal{L}^d(B)}.$$

In order to introduce the mean covariogram of the set $X \cap B$, let us denote $E_r = (X \cap B) \cap (ru + (X \cap B))$. Clearly we have the following inclusions

$$E_r \subset X \cap (ru + X) \cap B \text{ and } [X \cap (ru + X) \cap B] \setminus E_r \subset B \setminus (B \cap (ru + B)).$$

Noting that $\mathcal{L}^d(B \setminus (B \cap (ru + B))) = g_B(0) - g_B(ru)$, we obtain

$$\frac{\gamma_{X \cap B}(ru)}{\mathcal{L}^d(B)} \leq \frac{\mathbb{E}(\mathcal{L}^d(X \cap (ru + X) \cap B))}{\mathcal{L}^d(B)} \leq \frac{\gamma_{X \cap B}(ru)}{\mathcal{L}^d(B)} + \frac{g_B(0) - g_B(ru)}{\mathcal{L}^d(B)}.$$

This yields both an upper and a lower bound of $\mathbb{P}(ru \in X, 0 \notin X)$. We have

$$\mathbb{P}(ru \in X, 0 \notin X) \leq \frac{\gamma_{X \cap B}(0) - \gamma_{X \cap B}(ru)}{\mathcal{L}^d(B)}.$$

By property 8 of Proposition 16 and the second inequality of Lemma 18,

$$\limsup_{r \rightarrow 0} \frac{1}{|r|} \mathbb{P}(ru \in X, 0 \notin X) \leq \frac{1}{2} \frac{\mathbb{E}(V_u(X \cap B))}{\mathcal{L}^d(B)} \leq \frac{1}{2} \theta_{V_u}(X) + \frac{1}{2} \frac{V_u(B)}{\mathcal{L}^d(B)}.$$

As for the lower bound,

$$\mathbb{P}(ru \in X, 0 \notin X) \geq \frac{\gamma_{X \cap B}(0) - \gamma_{X \cap B}(ru)}{\mathcal{L}^d(B)} - \frac{g_B(0) - g_B(ru)}{\mathcal{L}^d(B)}.$$

Again, by Proposition 16 and the first inequality of Lemma 18, we have

$$\liminf_{r \rightarrow 0} \frac{1}{|r|} \mathbb{P}(ru \in X, 0 \notin X) \geq \frac{1}{2} \frac{\mathbb{E}(V_u(X \cap B))}{\mathcal{L}^d(B)} - \frac{1}{2} \frac{V_u(B)}{\mathcal{L}^d(B)} \geq \frac{1}{2} \theta_{V_u}(X) - \frac{1}{2} \frac{V_u(B)}{\mathcal{L}^d(B)}.$$

The two established inequalities are true for any nonempty open ball B . Noting R the radius of B ,

$$\frac{V_u(B)}{\mathcal{L}^d(B)} = \frac{\omega_{d-1} R^{d-1}}{\omega_d R^d} = \frac{\omega_{d-1}}{\omega_d} \frac{1}{R}.$$

The enunciated formula is obtained by letting R tends to $+\infty$. \square

5.3 Computation of the specific variations of Boolean models

In this section we apply Theorem 17 to compute the specific directional variations and the specific variation of any stationary Boolean model. The Boolean model [27], [25] with intensity λ and grain distribution P_X is the stationary RACS Z defined by

$$Z = \bigcup_{i \in \mathbb{N}} x_i + X_i,$$

where $\{x_i, i \in \mathbb{N}\} \subset \mathbb{R}^d$ is a stationary Poisson point process with intensity $\lambda > 0$ and $(X_i)_{i \in \mathbb{N}}$ is a sequence of i.i.d. RACS with common distribution P_X , independent of $\{x_i, i \in \mathbb{N}\}$. Moreover, the RACS (X_i) are supposed to have a finite mean Lebesgue measure (otherwise $Z = \mathbb{R}^d$ a.s.). The avoiding functional of the Boolean model Z is well-known: for any compact $K \subset \mathbb{R}^d$ we have

$$\mathbb{P}(Z \cap K = \emptyset) = \exp(-\lambda \mathbb{E}(\mathcal{L}^d(X \oplus \check{K}))), \quad (8)$$

where X denotes a RACS with distribution P_X and $X \oplus \check{K} = \{x - y, x \in X, y \in K\}$ (see e.g. [27, p. 65] or [15, p. 164]). Starting from the general expression of Eq. (8) (which determines the distribution of Z), let us compute the variogram ν_Z of Z . For $K = \{0\}$, Eq. (8) becomes

$$q := \mathbb{P}(0 \notin Z) = \exp(-\lambda \mathbb{E}(\mathcal{L}^d(X))).$$

For $K = \{0, -ru\}$, with $r \neq 0$ and $u \in S^{d-1}$, remark that we have

$$\mathcal{L}^d(X \oplus \check{K}) = \mathcal{L}^d(X \cup ru + X) = 2\mathcal{L}^d(X) - \mathcal{L}^d(X \cap (ru + X)).$$

Hence in this case $\mathbb{E}(\mathcal{L}^d(X \oplus \check{K})) = 2\mathbb{E}(\mathcal{L}^d(X)) - \gamma_X(ru)$. As a result the variogram ν_Z is equal to [27, p. 68], [15, p. 165]

$$\begin{aligned} \nu_Z(ru) &= \mathbb{P}(-ru \in Z \text{ and } 0 \notin Z) = \mathbb{P}(0 \notin Z) - \mathbb{P}(Z \cap \{0, -ru\} = \emptyset) \\ &= q - \exp(-\lambda (2\mathbb{E}(\mathcal{L}^d(X)) - \gamma_X(ru))) \\ &= q - q \exp(-\lambda (\gamma_X(0) - \gamma_X(ru))). \end{aligned}$$

By Theorem 17 and property 8 of Proposition 16 we deduce

$$\theta_{V_u}(Z) = 2(\nu_Z^u)'(0+) = 2q\lambda \frac{1}{2} \mathbb{E}(V_u(X)) = \lambda \mathbb{E}(V_u(X)) \exp(-\lambda \mathbb{E}(\mathcal{L}^d(X))).$$

Integrating this formula over all directions u we obtain $\theta_V(Z)$. Our computation is summarized in the following statement.

Proposition 19 (Specific variations of a stationary Boolean model). *Let Z be the Boolean model with Poisson intensity λ and grain distribution P_X , let X be a RACS with distribution P_X , and suppose that $\mathbb{E}(\mathcal{L}^d(X)) < +\infty$. Then for all $u \in S^{d-1}$,*

$$\theta_{V_u}(Z) = \lambda \mathbb{E}(V_u(X)) \exp(-\lambda \mathbb{E}(\mathcal{L}^d(X)))$$

and

$$\theta_V(Z) = \lambda \mathbb{E}(\text{Per}(X)) \exp(-\lambda \mathbb{E}(\mathcal{L}^d(X))). \quad (9)$$

Eq. (9) is valid for any grain distribution P_X and generalizes known results for Boolean models with convex grains [25, p. 386]. Similar generalizations involving intensity of surface measures deriving from Steiner's formula have recently been established [13, 29]. As already stressed out, our result is similar but not identical since the outer Minkowski content of a set differs from its (variational) perimeter [28].

A promising direction for further works is to extend the notion of specific variation for non stationary RACS. In particular, following [29], one could try to derive local variation densities of certain non stationary Boolean models.

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