Limit theorems for excursion sets of stationary random fields

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Abstract We give an overview of the recent asymptotic results on the geometry of excursion sets of stationary random fields. Namely, we cover a number of limit theorems of central type for the volume of excursions of stationary (quasi-, positively or negatively) associated random fields with stochastically continuous realizations for a fixed excursion level. This class includes in particular Gaussian, Poisson shot noise, certain infinitely divisible, $\alpha$-stable and max-stable random fields satisfying some extra dependence conditions. Functional limit theorems (with the excursion level being an argument of the limiting Gaussian process) are reviewed as well. For stationary isotropic $C^1$-smooth Gaussian random fields similar results are available also for the surface area of the excursion set. Statistical tests of Gaussianity of a random field which are of importance to real data analysis as well as results for an increasing excursion level round up the paper.

1 Introduction

Geometric characteristics such as Minkowski functionals (or intrinsic volumes, curvature measures, etc.) of excursions of random fields are widely used for data analysis purposes in medicine (brain fMRI analysis, see e.g. [5], [55], [60], [62]), physics and cosmology (microwave background radiation analysis, see e.g. [41] and references therein), and materials science (quantification of porous media, see e.g. [42], [61]), to name just a few. Minkowski functionals include the volume, the surface area and the Euler–Poincaré characteristic (reflecting porosity) of a set with a sufficiently regular boundary.

Among the possible abundance of random field models, Gaussian random fields are best studied due to their analytic tractability. A number of results starting with

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explicit calculation of the moments of Minkowski functionals is available for them since the mid seventies of the last century. We briefly review these results in Section 4. However, our attention is focused on the asymptotic arguments for (mainly non-Gaussian) stationary random fields. There has been a recent breakthrough in this domain starting with the paper [15] where a central limit theorem (CLT) for the volume of excursions of a large class of quasi-associated random fields was proved. We also cover a number of hard-to-find results from recent preprints and PhD theses.

The paper is organized as follows. After introducing some basic facts on excursions and dependence structure of stationary random fields in Section 2, we briefly review the limit theorems for excursions of stationary Gaussian processes (\(d = 1\)) in the next section. However, our focus is on the recent results in the multidimensional case \(d > 1\) which is considered in Sections 5 and 6. Thus, Section 5 gives (uni- and multivariate as well as functional) central limit theorems for the volume of excursion sets of stationary (in general, non-Gaussian) random fields over fixed, variable or increasing excursion levels. In Section 6, a similar scope of results is covered for the surface area of the boundary of excursion sets of stationary (but possibly anisotropic) Gaussian random fields in different functional spaces. The paper concludes with a number of open problems.

2 Preliminaries

Fix a probability space \((\Omega, \mathcal{F}, P)\). Let \(X = \{X(t, \omega), t \in \mathbb{R}^d, \omega \in \Omega\}\) be a stationary (in the strict sense) real valued measurable (in \((t, \omega) \in \mathbb{R}^d \times \Omega\)) random field. Later on we suppress \(\omega\) in the notation. For integrable \(X\) we assume \(X\) to be centered (i.e., \(E_X(o) = 0\) where \(o \in \mathbb{R}^d\) is the origin point). If the second moment of \(X(o)\) exists then we denote by \(C(t) = E(X(o)X(t)), t \in \mathbb{R}^d\) the covariance function of \(X\).

Let \(\| \cdot \|_2\) be the Euclidean norm in \(\mathbb{R}^d\) and \(\text{dist}_2\) the Euclidean distance: for two sets \(A, B \subset \mathbb{R}^d\) we put \(\text{dist}_2(A, B) = \inf\{\|x - y\|_2 : x \in A, y \in B\}\). Denote by \(\| \cdot \|_\infty\) the supremum norm in \(\mathbb{R}^d\) and by \(\text{dist}_\infty\) the corresponding distance function.

Let \(\varnothing \rightarrow\) mean convergence in distribution. Denote by \(A^c\) the complement and by \(\text{int}(A)\) the interior of a set \(A\) in the corresponding ambient space which will be clear from the context. Let \(\text{card}(A)\) be the cardinality of a finite set \(A\). Denote by \(B_r(x)\) the closed Euclidean ball with center in \(x \in \mathbb{R}^d\) and radius \(r > 0\). Let \(\mathcal{H}^k(\cdot)\) be the \(k\)-dimensional Hausdorff measure in \(\mathbb{R}^d\), \(0 \leq k \leq d\). In the sequel, we use the notation \(\kappa_j = \mathcal{H}^j(B_1(o)), j = 0, \ldots, d\).

To state limit theorems, one has to specify the way of expansion of windows \(W_n \subset T\), where the random field \(X = \{X(t), t \in T\}\) is observed, to the whole index space \(T = \mathbb{R}^d\) or \(\mathbb{Z}^d\). A sequence of compact Borel sets \((W_n)_{n \in \mathbb{N}}\) is called a Van Hove sequence (VH) if \(W_n \uparrow \mathbb{R}^d\) with

\[
\lim_{n \to \infty} V_d(W_n) = \infty \quad \text{and} \quad \lim_{n \to \infty} \frac{V_d(\partial W_n \oplus B_r(o))}{V_d(W_n)} = 0, \quad r > 0.
\]
A sequence of finite subsets $U_n \subset \mathbb{Z}^d$, $n \in \mathbb{N}$ is called regular growing if

$$\text{card}(U_n) \to \infty \quad \text{and} \quad \text{card}(\partial U_n)/\text{card}(U_n) \to 0 \quad \text{as} \quad n \to \infty$$

where $\partial U_n = \{j \in \mathbb{Z}^d \setminus U_n : \text{dist}_u(j, U_n) = 1\}$ is the discrete boundary of $U_n$ in $\mathbb{Z}^d$.

### 2.1 Excursion sets and their intrinsic volumes

The excursion set of $X$ at level $u \in \mathbb{R}$ in the compact observation window $W \subset \mathbb{R}^d$ is given by $A_u(X, W) = \{t \in W : X(t) \geq u\}$. The sojourn set under the level $u$ is $S_u(X, W) = \{t \in W : X(t) \leq u\}$, respectively.

Due to measurability of $X$, $A_u(X, W)$ and $S_u(X, W)$ are random Borel sets. If $X$ is a.s. upper (lower) semicontinuous then $A_u(X, W)$ ($S_u(X, W)$, respectively) is a random closed set (cf. [45, Section 5.2.1]).

A popular way to describe the geometry of excursion sets is via their intrinsic volumes $V_j$, $j = 0, \ldots, d$. They can be introduced for various families of sets such as convex and polyconvex sets [54, Chapter 4], sets of positive reach and their finite unions [22], unions of basic complexes [4, Chapter 6]. One possibility to define $V_j(K)$, $j = 0, \ldots, d$ for a set $K$ belonging to the corresponding family is given by the Steiner formula (see e.g. [53, Section 13.3]) as the coefficients in the polynomial expansion of the volume of the tubular neighbourhood $K_r = \{x \in \mathbb{R}^d : \text{dist}_2(x, K) \leq r\}$ of $K$ with respect to the radius $r > 0$ of this neighbourhood:

$$H^d(K_r) = \sum_{j=0}^{d} \kappa_{d-j} V_j(K) r^{d-j}$$

for admissible $r > 0$ (for convex $K$, these are all positive $r$). The geometric interpretation of intrinsic volumes $V_j(K)$, $j = 1, \ldots, d - 2$ can be given in terms of integrals of elementary symmetric polynomials of principal curvatures for convex sets $K$ with $C^2$–smooth boundary, cf. [53, Sections 13.5-6]. Without going into details here, let us discuss the meaning of some of $V_j(A_u(X, W))$, $j = 0, \ldots, d$ in several dimensions.

For $d = 1$, $V_1(A_u(X, W))$ is the length of excursion intervals and $V_0(A_u(X, W))$ is the number of upcrossings of level $u$ by the random process $X$ within $W$.

For dimensions $d \geq 2$, $V_d(A_u(X, W))$ is always the volume (i.e., the Lebesgue measure) of $A_u(X, W)$ and $V_{d-1}(A_u(X, W))$ is half the surface area, i.e., $1/2 \cdot H^{d-1}(\partial A_u(X, W))$. The Euler characteristic $V_0(A_u(X, W))$ is a topological measure of “porosity” of excursion set $A_u(X, W)$. For “basic” sets $A$ (e.g., non–empty convex sets or sets of positive reach) we set $V_0(A) = 1$. Then $V_0$ is defined for unions of basic sets by additivity. One can show that for $d = 2$ it holds

$$V_0(A) = \text{card}\{\text{connented components of } A\} - \text{card}\{\text{holes of } A\}.$$
of lower orders \( j = 0, \ldots, d - 2 \) are well defined e.g. for excursion sets of sufficiently smooth (at least \( C^2 \)) deterministic functions (cf. [4] Theorem 6.2.2) and Gaussian random fields (cf. [4] Theorem 11.3.3) satisfying some additional conditions.

### 2.2 Dependence concepts for random fields

To prove limit theorems for a random field \( X \), some conditions have to be imposed on the structure of the dependence of \( X \). Mixing conditions that are usually required (cf. e.g. [20], [13]) are however rather difficult to check for a particular random field under consideration. For this practical reason, we follow the books [15], [58, Chapter 10] and introduce association as well as related dependence concepts.

A random field \( X = \{ X(t), t \in \mathbb{R}^d \} \) is called associated if

\[
\text{cov}(f(X_I), g(X_J)) \geq 0
\]

for any finite subset \( I \subset \mathbb{R}^d \), and for any bounded coordinatewise non-decreasing functions \( f : \mathbb{R}^{\text{card}(I)} \to \mathbb{R}, g : \mathbb{R}^{\text{card}(J)} \to \mathbb{R} \) where \( X_I = \{ X(t), t \in I \} \).

A random field \( X = \{ X(t), t \in \mathbb{R}^d \} \) is called positively (PA) or negatively (NA) associated if

\[
\text{cov}(f(X_I), g(X_J)) \geq 0 \quad (\leq 0, \text{ resp.})
\]

for all finite disjoint subsets \( I, J \subset \mathbb{R}^d \), and for any bounded coordinatewise non-decreasing functions \( f : \mathbb{R}^{\text{card}(I)} \to \mathbb{R}, g : \mathbb{R}^{\text{card}(J)} \to \mathbb{R} \). It is clear that if \( X \in \text{A} \) then \( X \in \text{PA} \).

Subclasses of \( \text{A} (\text{PA}, \text{NA}) \)- fields are certain infinitely divisible (e.g., max-stable and \( \alpha \)-stable) random fields. In particular, a Gaussian random field with non-negative covariance function is associated.

A random field \( X = \{ X(t), t \in \mathbb{R}^d \} \) with finite second moments is called quasi-associated (QA) if

\[
|\text{cov}(f(X_I), g(X_J))| \leq \sum_{i \in I} \sum_{j \in J} \text{Lip}_i(f) \text{Lip}_j(g) |\text{cov}(X(i),X(j))|
\]

for all finite disjoint subsets \( I, J \subset \mathbb{R}^d \), and for any Lipschitz functions \( f : \mathbb{R}^{\text{card}(I)} \to \mathbb{R}, g : \mathbb{R}^{\text{card}(J)} \to \mathbb{R} \) where \( \text{Lip}_i(f) \) is the Lipschitz constant of function \( f \) for coordinate \( i \). It is known that if square integrable \( X \in \text{A(PA,NA)} \) then \( X \in \text{QA} \), cf. [15, Theorem 5.3].

A real-valued random field \( X = \{ X(t), t \in \mathbb{R}^d \} \) is called \( (BL, \theta) \)-dependent if there exists a non-increasing sequence \( \theta = \{ \theta_r \}_{r \in \mathbb{Q}^+} \), \( \theta_r \downarrow 0 \) as \( r \to \infty \) such that for any finite disjoint sets \( I, J \subset \mathbb{R}^d \) with \( \text{dist}_r(I,J) = \inf_{r \in \mathbb{Q}^+} r \), and any bounded Lipschitz functions \( f : \mathbb{R}^{\text{card}(I)} \to \mathbb{R}, g : \mathbb{R}^{\text{card}(J)} \to \mathbb{R} \), one has

\[
|\text{cov}(f(X_I), g(X_J))| \leq \sum_{i \in I} \sum_{j \in J} \text{Lip}_i(f) \text{Lip}_j(g) |\text{cov}(X(i),X(j))| \theta_r.
\]
It is often possible to choose \( \theta \) as the Cox–Grimmett coefficient

\[
\theta_r = \sup_{y \in \mathbb{R}^d \setminus B_r(y)} \int_{\mathbb{R}^d \setminus B_r(y)} |\text{cov}(X(y), X(t))| \, dt
\]

where \( B^r_r(y) = \{ x \in \mathbb{R}^d : \|x - y\|_\infty \leq r \} \). It can be easily seen that if \( X \in QA \) and its covariance function is absolutely integrable on \( \mathbb{R}^d \) then \( X \) is \((BL, \theta)\)-dependent.

## 3 Excursions of stationary Gaussian processes

Excursions of stochastic processes is a popular research topic in probability theory since many years, see e.g. [10] and references in [27]. The vast literature on this subject for different classes of processes such as Lévy, diffusion, stable, Gaussian ones, etc. can be hardly covered by one review. For this reason, we concentrate on the excursions of (mainly stationary) Gaussian processes here.

Let \( X = \{X(t), t \geq 0\} \) be a centered real valued Gaussian process. If \( X \) is a polynomial of degree \( n \) with iid \( N(0, 1) \)-distributed coefficients then the mean number of real roots of the equation \( X(t) = 0 \) was first obtained by M. Kac [28]. It initiated a substantial amount of papers on the roots of random algebraic polynomials, see [12] for a review. For \( C^1 \)-smooth stationary Gaussian processes \( X \), expectation of the number of upcrossings of a level \( u \) by \( X \) in time interval \([0, 1]\) has been studied in [50, 51], etc. Higher order factorial moments are considered in [17], see also references therein, and [7, 8]. For reviews (also including results on non–Gaussian stationary processes) see [33, Sections 7.2-7.3] and [6, Chapter 3]. In [1] and [2], the notion of the number of upcrossings of level \( u \) for random processes has been generalized to the Euler-Poincaré characteristic of excursion sets of random fields.

The first proof of a central limit theorem for the number of zeros of a stationary Gaussian process within an increasing time interval was given in [40]. Cuzick [18] refined the assumptions given in [40] and proved a central limit theorem for the number of zeros \( N_X(T) = 2V_0(A_0(X; [0, T])) \) of a centered separable stationary Gaussian process \( X = \{X(t), t \geq 0\} \) in the time interval \([0, T]\) as well as analogous results for integrals \( \int_0^T g(X(t)) \, dt \) as \( T \to \infty \). He used approximations by \( m \)-dependent random processes with spectral representation as a method borrowed from [40]. In more detail, let \( C(t) \) be twice differentiable with \( C(0) = 1, C''(0) = -\lambda_2 \) and variogram \( \gamma \) of \( X' \) be given by \( \gamma(h) = C''(h) - C''(0) = 1/2E(X'(h) - X'(0)), h \geq 0 \).

**Theorem 1** ([18]). If \( C, C'' \) are square integrable on \( \mathbb{R}_+ \), \( \int_0^T \gamma(t)/t \, dt < \infty \) for some \( \varepsilon > 0 \) and

\[
\text{Var}N_X(T)/T \to \sigma^2_2 > 0 \text{ as } T \to +\infty
\]

then

\[
T^{-1/2}(N_X(T) - \mathbb{E}N_X(T)) \xrightarrow{d} N(0, \sigma^2_2) \text{ as } T \to +\infty
\]

where
\[ \sigma^2 = \pi^{-1} \left( \lambda_2^{1/2} + \int_0^\infty \left( \frac{E(|X'(0)X'(t)||X(0) = X(t) = 0) - (E|X'(0)|^2)}{\sqrt{1 - C^2(t)}} \right) dt \right). \]

Condition (1) is difficult to check and is substituted in [18, Lemma 5] by a more tractable sufficient condition involving \( C \) and \( \lambda_2 \). Piterbarg [48] managed to prove the above theorem by substituting condition (1) with

\[ \int_0^\infty t \left( |C(t)| + |C'(t)| + |C''(t)| \right) dt < \infty. \]

He approximates the point process of upcrossings of \( X \) of level \( u \) by a strongly mixing point process.

**Theorem 2 ([18]).** Let \( X \) be a stationary Gaussian process with covariance function \( C \) being integrable on \( \mathbb{R}_+ \). For any measurable function \( g : \mathbb{R} \to \mathbb{R} \) such that \( Eg^2(X(0)) < \infty \) and \( g(x) - g(0) \) is not odd it holds

\[ T^{-1/2} \left( \int_0^T g(X(t)) dt - TEg(X(0)) \right) \xrightarrow{d} N(0, \sigma^2) \text{ as } T \to +\infty \quad (2) \]

where \( \sigma^2 > 0 \).

It is clear that the choice \( g(x) = \mathbf{1}\{x \in \mathbb{R} : x \geq u\} \) for any \( u \in \mathbb{R} \) leads to the central limit theorem for the length \( V_1(A_u(X;[0,T])) \) of excursion intervals of \( X \) in \([0,T]\).

Elizarov [21] first proved a functional central limit theorem for the sojourn times of the stationary Gaussian process under the level \( u \), in our terms, for \( V_1(S_u(X;[0,T])) \) if excursion level \( u \) is allowed to vary within \( \mathbb{R} \). Additionally, an analogous result for local times

\[ \lim_{\varepsilon \to 0} \frac{1}{2\varepsilon} (V_1(S_{u+\varepsilon}(X;[0,T])) - V_1(S_{u-\varepsilon}(X;[0,T]))) \]

was given. Both results were proved in the functional space \( C[0,1] \) after the substitution \( u \mapsto f(x), x \in [0,1] \) where \( f \in C[0,1] \) is a monotonously increasing function with \( f(0) = -\infty, f(1) = \infty \).

Belyaev and Nosko [9] proved limit theorems for \( V_1(A_u(X;[0,T])) \), \( T \to \infty \) as \( u \to \infty \) for stationary ergodic processes \( X \) satisfying a number of additional (quite technical) assumptions. In particular, these assumptions are satisfied if \( X \) is an ergodic Gaussian stationary process with twice continuously differentiable covariance function such that

\[ |C''(t) - C''(0)| \leq a/|\log |t||^{1+\varepsilon}, \quad t \downarrow 0 \]

for some constants \( a, \varepsilon > 0 \).

Slud [57] gave a multiple Wiener-Itô representation for the number of crossings of a \( C^1 \)-function \( \psi \) by \( X \). In [31], methods of [40] and [18] are generalized to the
case of functionals of $X, X'$ and $X''$. CLTs for the number of crossings of a smooth curve $\psi$ by a Gaussian process $X$ as well as for the number of specular points of $X$ (if $X$ is a Gaussian process in time and space) are given in [32]. For a review of results on moments and limit theorems for different characteristics of stationary Gaussian processes see [30]. In [27], CLTs for the multivariate non-linear weighted functionals (similar to those in (2)) of Gaussian stationary processes with multiple singularities in their spectra, having a covariance function belonging to a certain parametric family, are proved.

4 Moments of $V_j(A_u(X, W))$ for Gaussian random fields

We briefly review the state of the art for $\mathbb{E}V_j(A_u(X, W))$ of Gaussian random fields $X$. For recent extended surveys see the books [4] and [6]. For stationary (isotropic) Gaussian fields $X$, stratified $C^2$-smooth compact manifolds $W \subset \mathbb{R}^d$ and any $u \in \mathbb{R}$, formulae for $\mathbb{E}V_j(A_u(X, W))$, $j = 0, \ldots, d$ are given in [4] Theorems 13.2.1 and 13.4.1.

Apart from obtaining exact (or asymptotic as $u \to \infty$) formulae for $\mathbb{E}V_j(A_u(X, W))$, $j = 0, \ldots, d$, the possibility of an estimate

$$\mathbb{P}\left(\sup_{t \in W} X(t) > u\right) - \mathbb{E}V_0(A_u(X, W)) \leq g(u)$$

(the so-called Euler-Poincaré heuristic) with $g(u) = o(1)$ as $u \to \infty$ is of special interest. It has been proved in [3] Theorem 14.3.3 with $g(u) = c_0 \exp\{-u^2(1 + \alpha)/2\}$ for some positive constants $c_0$ and $\alpha$ if $X$ is a (non)stationary Gaussian random field with constant variance on a stratified manifold $W$ as $u \to \infty$. Lower and upper bounds for the density of supremum of stationary Gaussian random fields $X$ (which imply relation (3)) for any $u \in \mathbb{R}$ are given in [6] Theorem 8.4. Similar bounds are proven in [6] Theorem 8.10] for non-stationary Gaussian random fields $X$ with a unique point of maximum of variance in int($W$) as $u \to \infty$.

In [59], asymptotic behavior of $\mathbb{E}V_j(A_u(X, [a, b]^d))$, $j = 0, d - 1, d$ of non-stationary sufficiently smooth Gaussian random fields is studied as the excursion level $u \to \infty$. The variance of these fields is assumed to attain a global maximum at a vertex of $[a, b]^d$. It is shown that the heuristic (3) still holds true.

A interesting rather general formula for the mean surface area of Gaussian excursions is proven in [24]. Let $W$ be a compact subset of $\mathbb{R}^d$ with a non-empty interior and a finite Hausdorff measure of the boundary. Let $X = \{X(t), t \in W\}$ be a Gaussian random field with mean $\mu(t) = \mathbb{E}X(t)$ and variance $\sigma^2(t) = \text{Var}X(t)$. For an arbitrary (but fixed) excursion level $u \in \mathbb{R}$ introduce the zero set $\nabla_X^{-1}(0)$ of the gradient of the normalized field $(X - u)/\sigma$ by $\nabla_X^{-1}(0) = \{t \in W : \nabla((X(t) - u)/\sigma(t)) = 0\}$.

**Theorem 3** ([24]). Assume that $X \in C^1(W)$ a.s., $\mathbb{E}V_{d-1}\left(\nabla_X^{-1}(0)\right) < \infty$ and $\sigma(t) > 0$ for all $t \in W$. Then
$\mathbb{EV}_{d-1}(\partial A_u(X,W)) = \frac{1}{2\sqrt{2\pi}} \int_W \exp \left[ -\frac{(\mu(t) - u)^2}{2\sigma^2(t)} \right] \mathbb{E} \left\| \nabla ((X(t) - u)/\sigma(t)) \right\| dt.$

Asymptotic formulae for $\mathbb{EV}_j(A_u(X,W))$, $j = 0, \ldots, d$ as $u \to \infty$ of three subclasses of stable random fields (subgaussian, harmonizable, concatenated–harmonizable ones) are given in [3].

## 5 Volume of excursion sets of stationary random fields

The first limit theorems of central type for the volume of excursion sets (over a fixed level $u$) of stationary isotropic Gaussian random fields were proved in [26, Chapter 2]. There, the case of short and long range dependence (Theorem 2.2.4 and Example 2.2.1, Theorem 2.4.6) was considered. The CLT followed from a general Berry-Esséen-type bound for the distribution function of properly normed integral functionals

$$\int_{B_r(o)} G(X(t)) dt$$

as $r \to \infty$ where $G : \mathbb{R} \to \mathbb{R}$ is a function such that $\mathbb{E}G^2(X(o)) < \infty$ satisfying some additional assumptions, cf. also [36]. To get the volume $\mathbb{V}_d(A_u(X,B_r(o)))$ out of (4), set $G(x) = 1(|x| \geq u)$. The isotropy of $X$ was essential as one used expansions with respect to the basis of Chebyshev-Hermite polynomials in the proofs. The cases of $G(x) = 1(|x| \geq u)$, $\max \{0, x\}, |x|$ as well as of $G$ depending on a parameter and of weighted integrals in (4) are considered as well.

In a remark [26, p. 81], it was noticed that similar CLTs can be expected for non-Gaussian mixing random fields. The aim of this Section is to review the recent advances in proving such CLTs for various classes of stationary random fields that include also the (not necessarily isotropic) Gaussian case.

For instance, random fields with singularities of their spectral densities are considered in [37]. In Section 3.2 of that book, non–central limit theorems for the volume of excursions of stationary isotropic Gamma correlated and $\chi^2$-random fields over a radial surface (i.e., the level $u$ is not constant anymore, but a function of $||t||_2$, where $t \in \mathbb{R}^d$ is the integration variable in (4)) are proved. (Non)central limit theorems for functionals (4) of stationary isotropic vector–valued Gaussian random fields are given in the recent preprint [34]. There, the case of long and short range dependence is considered as well as applications to $F$– and $t$–distributed random fields.

The asymptotic behavior of tail probabilities

$$P \left( \int_W e^{X(t)} dt > x \right), \quad x \to \infty$$
for a homogeneous smooth Gaussian random field $X$ on a compact $W \subset \mathbb{R}^d$ is considered in [38], see [39] for further extensions.

### 5.1 Limit theorems for a fixed excursion level

The main result (which we call a methatheorem) can be formulated as follows.

**Theorem 4 (Methatheorem).** Let $X$ be a strictly stationary random field satisfying some additional conditions and $u \in \mathbb{R}$ fixed. Then, for any sequence of $VH$-growing sets $W_n \subset \mathbb{R}^d$, one has

$$\frac{V_d(A_u(X,W_n)) - P(X(o) \geq u) \cdot V_d(W_n)}{\sqrt{V_d(W_n)}} \overset{d}{\to} \mathcal{N}(0, \sigma^2(u))$$

as $n \to \infty$. Here

$$\sigma^2(u) = \int_{\mathbb{R}^d} \text{cov}(1\{X(o) \geq u\}, 1\{X(t) \geq u\}) \, dt.$$  

Depending on the class of random fields, these additional conditions will vary. First we consider the family of square–integrable random fields.

#### 5.1.1 Quasi-associated random fields

**Theorem 5 ([15]).** Let $X = \{X(t), t \in \mathbb{R}^d\} \in QA$ be a stationary square-integrable random field with a continuous covariance function $C$ such that $|C(t)| = O(||t||^{-\alpha}_2)$ for some $\alpha > 3d$ as $||t||_2 \to \infty$. Let $X(o)$ have a bounded density. Then $\sigma^2(u) \in (0, \infty)$ and Theorem 4 holds true.

Let us give an idea of the proof. Introduce the random field $Z = \{Z(j), j \in \mathbb{Z}^d\}$ by

$$Z(j) = \int_{j+ [0,1]^d} 1\{X(t) \geq u\} \, dt - \Psi(u), \quad j \in \mathbb{Z}^d.$$  

Here $\Psi(u) = P(X(o) > u)$ is the tail distribution function of $X(o)$. It is clear that the sum of $Z(j)$ over indices $j \in W_n \cap \mathbb{Z}^d$ approximates the numerator in (5). One has to show that $Z$ can be approximated by a sequence of $(BL, \theta)$-dependent stationary centered square-integrable random fields $Z_\gamma$, $\gamma \downarrow 0$, on $\mathbb{Z}^d$. The proof finishes by applying the following CLT to $Z_\gamma$ for each $\gamma > 0$.

**Theorem 6 ([16], Theorem 3.1.12).** Let $Z = \{Z(j), j \in \mathbb{Z}^d\}$ be a $(BL, \theta)$-dependent strictly stationary centered square-integrable random field. Then, for any sequence of regularly growing sets $U_n \subset \mathbb{Z}^d$, one has

$$\frac{S(U_n)}{\sqrt{\text{card}(U_n)}} \overset{d}{\to} \mathcal{N}(0, \sigma^2)$$

where $S(U_n)$ denotes the sum over $U_n$. 

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LTs for excursion sets of random fields 9
as } n \to \infty, \text{ with } \\
\sigma^2 = \sum_{j \in \mathbb{Z}^d} \text{cov}(Z(o), Z(j)).

We give two examples of random fields satisfying Theorem 5.

Example 1 ([15]). Let } X = \{X(t), t \in \mathbb{R}^d\} \text{ be a stationary shot noise random field given by } X(t) = \sum_{\iota \in \mathbb{N}} \xi_{\iota} \phi(t - x_{\iota}) \text{ where } \Pi_{\lambda} = \{x_{\iota}\} \text{ is a homogeneous Poisson point process in } \mathbb{R}^d \text{ with intensity } \lambda \in (0, \infty), \{\xi_{\iota}\} \text{ is a family of i.i.d. non-negative random variables with } \mathbb{E}\xi_{\iota}^2 < \infty \text{ and characteristic function } \varphi_{\xi_{\iota}}. \text{ Assume that } \Pi_{\lambda} \text{ and } \{\xi_{\iota}\} \text{ are independent. Moreover, let } \varphi : \mathbb{R}^d \to \mathbb{R}^+ \text{ be a bounded and uniformly continuous Borel function with } \varphi(t) \leq g_0(\|t\|_2) = O(\|t\|_2^{-\alpha}) \text{ as } \|t\|_2 \to \infty \text{ for a function } g_0 : \mathbb{R}_+ \to \mathbb{R}_+, \alpha > 3d, \text{ and}

\int_{\mathbb{R}^d} \left| \exp \left\{ \lambda \int_{\mathbb{R}^d} \left( \varphi_{\xi_{\iota}}(s\varphi(t)) - 1 \right) dt \right\} \right| ds < \infty.

Then Theorem 5 holds true.

Example 2 ([15]). Consider a stationary Gaussian random field } X = \{X(t), t \in \mathbb{R}^d\} \text{ with a continuous covariance function } C(\cdot) \text{ such that } |C(t)| = O(\|t\|^{-\alpha}) \text{ for some } \alpha > d \text{ as } \|t\|_2 \to \infty. \text{ Let } X(o) \sim \mathcal{N}(a, \tau^2). \text{ Then, Theorem 5 holds true with}

\sigma^2(u) = \frac{1}{2\pi} \int_{\mathbb{R}^d} \int_0^{\rho(t)} \frac{1}{\sqrt{1 - s^2}} e^{-\frac{(u - a)^2}{2(1 + s^2)}} ds dt,

\text{where } \rho(t) = \text{corr}(X(o), X(t)). \text{ In particular, for } u = a \text{ one has}

\sigma^2(a) = \frac{1}{2\pi} \int_{\mathbb{R}^d} \arcsin(\rho(t)) dt.

5.1.2 PA- or NA-random fields

What happens if the field } X \text{ does not have the finite second moment? In this case, another set of conditions for our metatheorem to hold was proven in [29] Theorem 3.59.

Theorem 7. Let } X = \{X(t), t \in \mathbb{R}^d\} \in \text{PA(NA)} \text{ be stochastically continuous satisfying the following properties:
1. the asymptotic variance } \sigma^2(u) \in (0, \infty) \text{ (cf. its definition in (6)},
2. } P(X(o) = u) = 0 \text{ for the chosen level } u \in \mathbb{R}.

Then Theorem 4 holds.

The idea of the proof is first to show that the random field } Z = \{Z(j), j \in \mathbb{Z}^d\} \text{ defined in (12) is PA (NA). Second, use [16] Theorem 1.5.17] to prove that } Z \text{ is } (BL, \theta)-\text{dependent. Then apply Theorem 6 to } Z.
A number of important classes of random fields satisfy Theorem [7]. For instance, stationary infinitely divisible random fields $X = \{X(t), t \in \mathbb{R}^d\}$ with spectral representation

$$X(t) = \int_E f_t(x) \Lambda(dx), \quad t \in \mathbb{R}^d,$$

where $\Lambda$ is a centered independently scattered infinitely divisible random measure on space $E$ and $f_t : E \to \mathbb{R}_+$ are $\Lambda$-integrable kernels, are associated and hence PA by [16, Chapter 1, Theorem 3.27]. The finite susceptibility condition $\sigma^2(u) \in (0, \infty)$ can be verified by [29, Lemma 3.71]. Further examples of random fields satisfying Theorem [7] are stable random fields which we consider in more detail following [29, Section 3.5.3].

Max-stable random fields

Let $X = \{X(t), t \in \mathbb{R}^d\}$ be a stationary max-stable random field with spectral representation

$$X(t) = \max_{i \in \mathbb{N}} \xi_i f_i(x), \quad t \in \mathbb{R}^d,$$

where $f_i : E \to \mathbb{R}_+$ is a measurable function defined on the measurable space $(E, \mu)$ for all $t \in \mathbb{R}^d$ with

$$\int_E f_i(y) \mu(dy) = 1, \quad t \in \mathbb{R}^d,$$

and $\{(\xi_i, y_i)\}_{i \in \mathbb{N}}$ is the Poisson point process on $(0, \infty) \times E$ with intensity measure $\xi^{-2}d\xi \times \mu(dy)$. It is known that all max-stable distributions are associated and hence PA by [49, Proposition 5.5.29]. The field $X$ is stochastically continuous if $\|f_s - f_t\|_{L_1} \to 0$ as $s \to t$ (cf. [23, Lemma 2]). Condition $\sigma^2(u) \in (0, \infty)$ is satisfied if

$$\int_{\mathbb{R}^d} \int_E \min\{f_0(y), f_t(y)\} \mu(dy) dt < \infty.$$

$\alpha$-stable random fields

Let $X = \{X(t), t \in \mathbb{R}^d\}$ be a stationary $\alpha$-stable random field ($\alpha \in (0, 2)$, for simplicity $\alpha \neq 1$) with spectral representation

$$X(t) = \int_E f_t(x) \Lambda(dx), \quad t \in \mathbb{R}^d,$$

where $\Lambda$ is a centered independently scattered $\alpha$-stable random measure on space $E$ with control measure $m$ and skewness intensity $\beta : E \to [-1, 1]$, $f_t : E \to \mathbb{R}_+$ is a measurable function on $(E, m)$ for all $t \in \mathbb{R}^d$. By [52, Proposition 3.5.1], $X$ is stochastically continuous if $f_t |f_s(x) - f_t(x)|^\alpha m(dx) \to 0$ as $s \to t$ for any $t \in \mathbb{R}^d$. Condition $\sigma^2(u) \in (0, \infty)$ is satisfied if
\[ \int_{\mathbb{R}^d} \left( \int_{E} \min\{ |f_0(x)|^\alpha, |f_t(x)|^\alpha \} m(dx) \right)^{1/(1+\alpha)} dt < \infty. \]

### 5.2 A multivariate central limit theorem

If a finite number of excursion levels \( u_k \in \mathbb{R}, k = 1, \ldots, r \) is considered simultaneously, a multivariate analogue of Theorem 4 can be proven. Introduce the notation

\[ S_u(W_n) = (V_d(A_{u_1}(X,W_n)), \ldots, V_d(A_{u_r}(X,W_n)))^\top, \quad \Psi(u) = (\Psi(u_1), \ldots, \Psi(u_r))^\top. \]

**Theorem 8** (\cite{15}, \cite{29}). Let \( X \) be the above random field satisfying Theorem 4. Then, for any sequence of \( \mathcal{VH} \)-growing sets \( W_n \subset \mathbb{R}^d \), one has

\[ V_d(W_n)^{-1/2} (S_u(W_n) - \Psi(u) V_d(W_n)) \xrightarrow{d} \mathcal{N}(0, \Sigma(u)) \]

as \( n \to \infty \). Here, \( \Sigma(u) = (\sigma_{lm}(u))^{r \times 1}_{l,m=1} \) with

\[ \sigma_{lm}(u) = \int_{\mathbb{R}^d} \text{cov} \left( 1 \{ X(0) \geq u_l \}, 1 \{ X(t) \geq u_m \} \right) dt. \]

If \( X \) is Gaussian as in Example 2, we have

\[ \sigma_{lm}(u) = \frac{1}{2\pi} \int_{\mathbb{R}^d} \int_0^{\rho(r)} \frac{1}{\sqrt{1-s^2}} \exp \left\{ - \frac{(u_1 - a)^2 - 2r(u_1 - a)(u_m - a) + (u_m - a)^2}{2\tau^2(1-s^2)} \right\} ds dt. \]

However, the explicit computation of the elements of matrix \( \Sigma \) for the majority of fields \( X \) (except for Gaussianity) seems to be a very complex task. In order to overcome this difficulty in statistical applications of the methatheorem to testing, the matrix \( \Sigma \) can be (weakly) consistently estimated from one observation of a stationary random field \( X \), see \cite{47}, \cite{58, Section 9.8.3} and references therein.

**Statistical version of the CLT and tests**

Let \( X \) be a random field satisfying Theorem 4, \( u_k \in \mathbb{R}, k = 1, \ldots, r \) and \( (W_n)_{n \in \mathbb{N}} \) be a sequence of \( \mathcal{VH} \)-growing sets. Let \( \hat{\Sigma}_n = (\hat{\sigma}_{lm})^{r \times 1}_{l,m=1} \) be a weakly consistent estimator for the nondegenerate asymptotic covariance matrix \( \Sigma(u) \), i.e., for any \( l, m = 1, \ldots, r \)

\[ \hat{\sigma}_{lm} \xrightarrow{p} \sigma_{lm}(u) \quad \text{as} \quad n \to \infty. \]

Then
\[ \tilde{C}_n^{-1/2} V_d(W_n)^{-1/2} (S_u(W_n) - \Psi(u) V_d(W_n)) \xrightarrow{d} \mathcal{N}(0, I). \]  

Based on the latter relation, an asymptotic test for the following hypotheses can be constructed:

\[ H_0 : X \text{ is a random field satisfying Theorem}^{[2]} \text{ with tail distribution function } \Psi(\cdot) \]

vs. \( H_1 \) : negation of \( H_0 \). As a test statistic, we use

\[ T_n = V_d(W_n)^{-1} (S_u(W_n) - \Psi(u) V_d(W_n)) \]

which is asymptotically \( \chi^2_n \)-distributed by continuous mapping theorem and relation \([8] \): \( T_n \xrightarrow{d} \chi^2_n \) as \( n \to \infty \). Hence, reject the null-hypothesis at a confidence level \( 1 - \nu \) if \( T_n > \chi^2_{\nu-1} \) where \( \chi^2_{\nu-1} \) is the \( (1 - \nu) \)-quantile of \( \chi^2_n \)-law.

### 5.3 Functional limit theorems

A natural generalization of multivariate CLTs is a functional CLT where the excursion level \( u \in \mathbb{R} \) is treated as a variable, which also appears as a ("time") index in the limiting Gaussian process. In order to state the main results, introduce the Skorokhod space \( D(\mathbb{R}) \) of càdlàg functions on \( \mathbb{R} \) endowed with the usual Skorokhod topology, cf. \([11] \) Section 12]. Denote by \( \Rightarrow \) the weak convergence in \( D(\mathbb{R}) \).

Define the stochastic processes \( Y_n = \{ Y_n(u), u \in \mathbb{R} \} \) by

\[ Y_n(u) = \frac{1}{n^{d/2}} \left( V_d(A_n(X, [0, n]^d)) - n^d \Psi(u) \right), \quad u \in \mathbb{R}. \]  

Introduce the condition

\[ (\ast) \quad \text{For any subset } T = \{ t_1, \ldots, t_k \} \subset \mathbb{R}^d \text{ and its partition } T = T_1 \cup T_2 \text{ there exist some constants } c(T), \gamma > 0 \text{ such that} 
\]

\[ \text{cov} \left( \prod_{t \in T_1} \phi_{a,b}(X(t)), \prod_{t \in T_2} \phi_{a,b}(X(t)) \right) \leq c(T) \left( 1 + \text{dist}_w(T_1, T_2) \right)^{-3d + \gamma}, \]

where \( \phi_{a,b}(x) = 1(a < x \leq b) - P(a < X(o) \leq b) \) for any real numbers \( a < b \).

The following functional CLT is proven in \([43] \) Theorem 1 and Lemma 1. \[ \text{Theorem 9. Let } X = \{ X(t), t \in \mathbb{R}^d \} \text{ be a real valued stationary random field with a.s. continuous sample paths and a bounded density of the distribution of } X(o). \]

\[ \text{Let condition } (\ast) \text{ and Theorem}^{[2]} \text{ be satisfied. Then } Y_n \Rightarrow Y \text{ as } n \to \infty \text{ where } Y = \{ Y(u), u \in \mathbb{R} \} \text{ is a centered Gaussian stochastic process with covariance function} 
\]

\[ C_Y(u,v) = \int_{\mathbb{R}^d} \text{cov} \left( 1\{ X(0) \geq u \}, 1\{ X(t) \geq v \} \right) dt, \quad u, v \in \mathbb{R}. \]
In particular, condition \((\star)\) is satisfied if \(X \in \mathcal{A}\) is square integrable with covariance function \(C\) that admits a bound

\[ |C(t)| \leq \zeta (1 + \|t\|_\infty)^{-\lambda} \]

for all \(t \in \mathbb{R}^d\) and some \(\zeta > 0\), \(\lambda > 9d\). The proofs are quite technical involving a Móricz bound for the moment of a supremum of (absolute values of) partial sums of random fields on \(\mathbb{Z}^d\), cf. [46, Theorem 2].

For max–stable random fields introduced in Section 5.1.2 condition \((\star)\) is satisfied if for any \(T = \{t_1, \ldots, t_k\} \subseteq \mathbb{R}^d\) and its partition \(T = T_1 \cup T_2\) there exist some constants \(c(T), \gamma > 0\) such that

\[
\int_{\mathbb{R}^d} \min \left\{ \max_{t \in T_1} f_t(y), \max_{t_j \in T_2} f_{t_j}(y) \right\} \mu(dy) \leq c(T) (1 + \text{dist}_\infty(T_1, T_2))^{-(3d+\gamma)}. \tag{10}
\]

For \(\alpha\)–stable moving averages, i.e., \(\alpha\)–stable random fields from Section 5.1.2 with \(f_t(\cdot) = f(t - \cdot)\) for any \(t \in \mathbb{R}^d\), condition (10) should be replaced by

\[
\left( \int_{\mathbb{R}^d} \min \left\{ \max_{t \in T_1} f(t - y), \max_{t_j \in T_2} f(t_j - y) \right\}^\alpha m(dy) \right)^{1/(1+\alpha)} \leq c(T) (1 + \text{dist}_\infty(T_1, T_2))^{-(3d+\gamma)}. \]

These results are proven (under slightly more general assumptions) in [29, Section 3.5.5] together with analogous conditions for infinitely divisible random fields (that are too lengthy to give them in a review paper) as well as examples of random fields satisfying them.

Theorem 9 together with the continuous mapping theorem can be used to test hypotheses of Section 5.2 with test statistic

\[ T_n = \sup_{u \in \mathbb{R}} Y_n(u) \]

if a large deviation result for the limiting Gaussian process \(Y\) is available, cf. [43, Corollary 1].

### 5.4 Limit theorem for an increasing excursion level

If the level \(u \to \infty\) one may also expect that a CLT for the volume of the corresponding excursion set holds, provided that a particular rate of convergence of \(r\) to infinity is chosen in accordance with the expansion rate of the observation window.

First results of this type were proven in [26, Theorems 2.7.1, 2.7.2, 2.8.1] for stationary isotropic Gaussian random fields with short or long range dependence. A
generalization to the case of stationary $\mathbf{PA}$-random fields is given in a recent preprint [19]:

**Theorem 10.** Let $X = \{X(t), \; t \in \mathbb{R}^d\} \in \mathbf{PA}$ be a stationary random field with a continuous covariance function $C$ such that $|C(t)| = \Theta(\|t\|_2^{-\alpha})$ for some $\alpha > 3d$ as $\|t\|_2 \to \infty$. Let $X(o)$ have a bounded density $p_{X(o)}$. Assume that the variance of $V_d(A_{u_n}(X, [0,n]^d))$ being equal to

$$\sigma_n^2 = \int_{[0,n]^d} \int_{[-x,n-x]^d} \text{cov}(1\{X(o) \geq u_n\}, 1\{X(t) \geq u_n\}) \, dt \, dx$$

satisfies

$$\sigma_n^2 \to \infty, \quad n \to \infty. \quad (11)$$

Introduce $\gamma(x) = \sup_{y \geq x} p_{X(o)}(y), \; x \in \mathbb{R}$. Choose a sequence of excursion levels $u_n \to \infty$ such that

$$\frac{n^d \gamma^{2/3}(u_n)}{\sigma_n^{2(\alpha+3)/3}} \to 0, \quad n \to \infty. \quad (12)$$

Then it holds

$$\frac{V_d(A_{u_n}(X, [0,n]^d)) - n^d \mathbb{P}(X(o) \geq u_n)}{\sigma_n} \mathcal{N}(0,1) \quad (13)$$

as $n \to \infty$. Conditions (11), (12) are checked in [19] explicitly for stationary (non-isotropic) Gaussian as well as shot noise random fields leading to quite tractable simple expressions. For instance, it suffices to choose $u_n = O(\sqrt{\log n}), \; n \to \infty$ in the Gaussian case.

Student and Fisher–Snedecor random fields are considered in the recent preprint [34, Section 7]. CLTs for spherical measures of excess

$$\int_{\partial B_r(o)} 1\{X(t) > u(r)\} \mathcal{H}^{d-1}(dt)$$

of a stationary Gaussian isotropic random field $X$ over the moving level $u(r) \to \infty, \; r \to \infty$ are proved in [37, Section 3.3]. For yet another type of geometric measures of excess over a moving level see [35].

### 6 Surface area of excursion sets of stationary Gaussian random fields

Limit theorems for $V_{d-1}(A_n(X, W_n))$ have been first proven for one fixed level $u$ and a stationary isotropic Gaussian random field $X$ in [31] in dimension $d = 2$. There, the expansion of $V_{d-1}(A_n(X, W_n))$ in Hermite polynomials is used. In higher dimen-
Let us focus on functional LTs for $V_{d-1}(\partial A_d(X,W))$ proven in [44] for the phase space $L^2(\mathbb{R}, \nu)$ (where $\nu$ is a standard Gaussian measure in $\mathbb{R}$) and in [56] for the phase space $C(\mathbb{R})$.

Let $X = \{X(t), t \in \mathbb{R}^d\}, d > 1$, be a centered stationary and isotropic Gaussian random field with a.s. $C^1$-smooth paths and covariance function $C \in C^2(\mathbb{R}^d)$ satisfying $C(o) = 1$ as well as

$$|C(t)| + \frac{1}{1-C(t)} \sum_{i,j=1}^{d} \left| \frac{\partial^2 C(t)}{\partial t_i \partial t_j} \right| < g(t)$$

for large $\|t\|_2$ (where $t = (t_1, \ldots, t_d)^\top$) and a bounded continuous function $g : \mathbb{R}^d \to \mathbb{R}_+$ such that $\lim_{\|t\|_2 \to \infty} g(t) = 0$ and

$$\int_{\mathbb{R}^d} \sqrt{g(t)} \, dt < \infty.$$  

Denote by $\nabla X(t)$ the gradient of $X(t)$. Assume that the $(2d + 2)$-dimensional random vector $(X(o), X(t), \nabla X(o), \nabla X(t))^\top$ is non-degenerate for all $t \in \mathbb{R}^d \setminus \{o\}$. Let $\lambda_2 = -\partial^2 C(o)/\partial t_i^2$.

Introduce the sequence of random processes $\{Y_n\}, n \in \mathbb{N}$ by

$$Y_n(u) = \frac{2\lambda_2^{d/2-1}}{n^{d/2}} \left( V_{d-1} \left( \partial A_{d-1}(X, [0,n]^d) \right) - EV_{d-1} \left( \partial A_{d-1}(X, [0,n]^d) \right) \right)$$

where $u \in \mathbb{R}$. They will be interpreted as random elements in $L^2(\mathbb{R}, \nu)$. Let $\rightarrow$ denote the weak convergence of random elements in $L^2(\mathbb{R}, \nu)$. Let

$$\kappa(t) = f(X(t)) \exp\left\{ -X(t)^2/2 \right\} \|\nabla X(t)\|_2, \quad t \in \mathbb{R}^d.$$  

**Theorem 11** ([44]). Under the above assumptions on $X$ and $C$, it holds $Y_n \rightarrow Y$ as $n \rightarrow \infty$ where $Y$ is a centered Gaussian random element in $L^2(\mathbb{R}, \nu)$ with covariance operator

$$\text{Var}(Y,f)_{L^2(\mathbb{R}, \nu)} = \frac{1}{2\pi} \int_{\mathbb{R}^d} \text{cov}(\kappa(o), \kappa(t)) \, dt, \quad f \in L^2(\mathbb{R}, \nu).$$
For $d \geq 3$, processes $Y_n$ have a continuous modification $\tilde{Y}_n$ if conditions on $X$ starting from (14) are replaced by the following ones:

1. Covariance function $C$ as well as all its first and second order derivatives belong to $L^1(\mathbb{R})$.
2. There exist $\tau \in (0, 1)$ and $\beta > 0$ such that for all $h \in [-\tau, \tau]$ and $e_h = (h, 0, 0, \ldots, 0)^T \in \mathbb{R}^d$ the determinant of the covariance matrix of the vector
   \[
   \left( X(o), X(e_h), \frac{\partial X(o)}{\partial t_1}, \frac{\partial X(e_h)}{\partial t_1} \right)^T
   \]
   is not less than $|h|^\beta$.

Let $\rightarrow$ denote the weak convergence of random elements in $C(\mathbb{R})$. Denote by $p_{X(t)}(p_{X(o), X(t)})$ the density of $X(t)$ ( $(X(o), X(t))^T$), $t \in \mathbb{R}^d$, respectively. Set
\[
H_t(u, v) = \mathbb{E}(\|\nabla X(o)\|_2 \|\nabla X(t)\|_2 | X(o) = u, X(t) = v), \quad u, v \in \mathbb{R}, \quad t \in \mathbb{R}^d.
\]
In definition (15), assume $\lambda = 1$.

**Theorem 12 ([56]).** Under the above assumptions on $X$ and $C$, it holds $\tilde{Y}_n \rightarrow Y$ as $n \rightarrow \infty$ for $d \geq 3$ where $Y$ is a centered Gaussian random process with covariance function
\[
cov(Y(u), Y(v)) = \int_{\mathbb{R}^d} \left( H_t(u, v)p_{X(o), X(t)}(u, v) - (\mathbb{E}\|\nabla X(o)\|_2)^2 p_{X(o)}(u)p_{X(t)}(v) \right) dt
\]
for $u, v \in \mathbb{R}$.

The case $d = 2$ is still open.

### 7 Open problems

It is a challenging problem to prove the whole spectrum of limit theorems for $V_j(A_u(X, W_n))$ of lower orders $j = 0, \ldots, d - 2$ for isotropic $C^2$-smooth stationary Gaussian random fields. Functional limit theorems and the case of increasing level $u \rightarrow \infty$ are therein of special interest. Further perspective of research is the generalisation of these (still hypothetic) results to non–Gaussian random fields.

Another open problem is to prove limit theorems for a large class of functionals of non–Gaussian stationary random fields that includes the volume of excursion sets. It is quite straightforward to do this for
\[
\int_{W_n} g(X(t)) dt
\]
for a measurable function $g : \mathbb{R} \to \mathbb{R}$ such that $E g^2(X(o)) < \infty$. For more general classes of functionals of the field $X$ and the observation window $W_n$ it is still terra incognita.

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