

Limit Theorems in Stochastic Geometry

Overview

- ▶ Introduction
- ▶ Specific intrinsic volumes of random closed sets
- ▶ Scan statistics for Lévy measures and point processes
- ▶ Open problems
 - ▶ Wiener sausage
 - ▶ Level (excursion) sets of stationary random fields

Introduction

In stochastic geometry:

Limit theorems for

- ▶ Random geometric graphs (Penrose, Yukich, ...)
- ▶ Random polytopes (Bárány, Buchta, Hug, Reitzner, Schneider, ...)
- ▶ Random closed sets (RACS), geometrical random fields and measures (Heinrich, Molchanov, S., Vitale, Weil, ...)

Introduction

Preliminaries

\mathcal{K}	family of non-empty compact convex sets (convex bodies) in \mathbb{R}^d
\mathcal{R}	$= \left\{ \bigcup_{i=1}^n K_i : K_i \in \mathcal{K}, i = 1, \dots, n, \forall n \right\}$ convex ring
\mathcal{S}	$= \left\{ K : K \cap W \in \mathcal{R}, \forall W \in \mathcal{K} \right\}$ extended convex ring
$B_r(a)$	ball with center in a and radius r
κ_j	volume of $B_1(o)$ in $\mathbb{R}^j, j = 0, \dots, d$
$K_1 \oplus K_2$	$= \bigcup_{x \in K_2} (K_1 + x)$ Minkowski addition
$K_1 \ominus K_2$	$= \bigcap_{x \in K_2} (K_1 + x)$ Minkowski subtraction
$ \cdot $	Lebesgue measure (volume)

Introduction

Limit theorem for random polytopes

(Bárány (1982); Reitzner (2003, 2005); Vu (2005))

For $K \in \mathcal{K}$ with $\partial K \in \mathcal{C}^3$, let $K_n = \text{conv}(X_1, \dots, X_n)$, where $X_i \sim U(K)$ are *iid* random points in K . Then, as $n \rightarrow \infty$, it holds

- ▶ $\mathbb{E} |K_n| = |K| - c_1(d, K)n^{-1/(d+1)}$,
- ▶ $\text{Var} |K_n| \sim c_2(K)n^{-(d+3)/(d+1)}$,
- ▶ $(|K_n| - \mathbb{E} |K_n|) / \sqrt{\text{Var} |K_n|} \xrightarrow{d} N(0, 1)$.

Introduction

LT for RACS w. r. to Minkowski addition (Weil (1982))

- ▶ **Hausdorff metric:** For two nonempty compacts $A, B \subset \mathbb{R}^d$
 $d_H(A, B) = \min \{r > 0 : A \subseteq B \oplus B_r(o), B \subseteq A \oplus B_r(o)\}.$
- ▶ **Norm of a set:** $\|A\| = \sup\{|x| : x \in A\}$
- ▶ **Support function:** for $A \in \mathcal{K}$, $s_A(u) = \sup\{u \cdot v : v \in A\}$, $u \in \mathbf{S}^{d-1}$.
- ▶ **Expectation of RACS (Aumann (1965)):** $\mathbb{E}\Xi = \text{convex set with support function } \mathbb{E}s_{\Xi}(\cdot)$

For a sequence of *iid* RACS $\Xi_i \stackrel{d}{=} \Xi$ with $\mathbb{E}\|\Xi\| < \infty$, it holds

$$\sqrt{n} d_H((\Xi_1 \oplus \dots \oplus \Xi_n)/n, \mathbb{E}\Xi) \xrightarrow{d} \sup_{u \in \mathbf{S}^{d-1}} X(u), \quad n \rightarrow \infty,$$

where X is the centered Gaussian process on \mathbf{S}^{d-1} with cov. f. $C(u, v) = \mathbb{E}[s_{\Xi}(u)s_{\Xi}(v)] - \mathbb{E}s_{\Xi}(u)\mathbb{E}s_{\Xi}(v)$, $u, v \in \mathbf{S}^{d-1}$.

Limit theorems for various characteristics of RACS

For the **volume fraction**, **specific boundary surface**, **number of connected components**, etc.:

- ▶ Baddeley (1980)
- ▶ Mase (1982)
- ▶ Heinrich (1993)
- ▶ Heinrich, Molchanov (1999)
- ▶ Böhm, Heinrich, Schmidt (2004)
- ▶ ...

For **all specific intrinsic volumes**:

- ▶ Pantle, Schmidt, S. (2006, 2009)

Intrinsic volumes

Steiner's formula in \mathbb{R}^d

- ▶ There exist functionals $V_j : \mathcal{K} \rightarrow [0, \infty)$, $j = 0, \dots, d$, (Minkowski functionals, quermassintegrals or intrinsic volumes) such that for any $r > 0$ and $K \in \mathcal{K}$ it holds

$$|K \oplus B_r(o)| = \sum_{j=0}^d r^{d-j} \kappa_{d-j} V_j(K).$$

- ▶ Functionals V_0, \dots, V_d are additive, motion invariant, monotone with respect to inclusion, and continuous with respect to d_H .

Intrinsic volumes

Theorem (Hadwiger (1957))

Let $F : \mathcal{K} \rightarrow \mathbb{R}$ be any additive, motion invariant and continuous functional. Then, F can be represented in the form

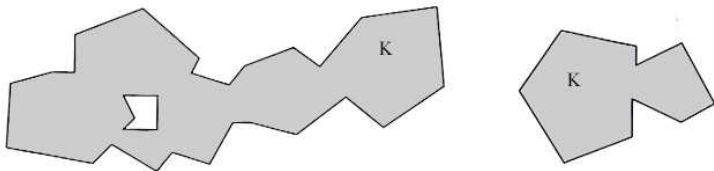
$$F = \sum_{j=0}^d a_j V_j$$

for some constants $a_0, \dots, a_d \in \mathbb{R}$.

Thus, the intrinsic volumes V_0, \dots, V_d form a **basis** in the corresponding linear space.

Intrinsic volumes

Additive extension to the convex ring \mathcal{R}



For each $j = 0, \dots, d$, there exists a unique additive extension of $V_j : \mathcal{K} \rightarrow [0, \infty)$ to \mathcal{R} given by the **inclusion–exclusion formula**:

$$V_j(K_1 \cup \dots \cup K_n) = \sum_{i=1}^n (-1)^{i-1} \sum_{j_1 < \dots < j_i} V_j(K_{j_1} \cap \dots \cap K_{j_i}), \quad K_1, \dots, K_n \in \mathcal{K}$$

Intrinsic volumes

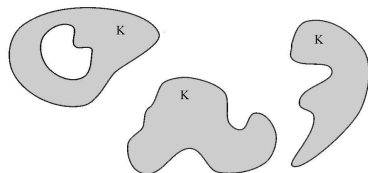
Geometrical interpretation: For any $K \in \mathcal{R}$ with $K \neq \emptyset$,

$$V_d(K) = |K| \quad (\text{volume})$$

$$2V_{d-1}(K) = \mathcal{H}^{d-1}(\partial K) \quad (\text{surface area})$$

$$V_0(K) = \chi(K) \quad (\text{Euler number})$$

In \mathbb{R}^2 : $\chi(K) = \#\{\text{clumps}\} - \#\{\text{holes}\}$



$$\chi(K) = 3 - 1 = 2$$

Random closed sets

- ▶ Let $(\Omega, \mathcal{F}, \mathcal{P})$ be an arbitrary probability space
- ▶ \mathfrak{C} = family of all compact sets in \mathbb{R}^d
- ▶ \mathfrak{F} = family of all closed sets in \mathbb{R}^d
- ▶ $\sigma(\mathfrak{F})$ = σ -algebra in \mathfrak{F} , generated by the sets $F_C = \{F \in \mathfrak{F} : F \cap C \neq \emptyset\}$ for any $C \in \mathfrak{C}$

An $(\mathcal{F}, \sigma(\mathfrak{F}))$ -measurable mapping $\Xi : \Omega \rightarrow \mathfrak{F}$ is called a **random closed set (RACS)**. Its distribution is uniquely determined by the **capacity functional** $T_\Xi(C) = P(\Xi \cap C \neq \emptyset)$, $C \in \mathfrak{C}$

Random closed sets

Stationarity and isotropy

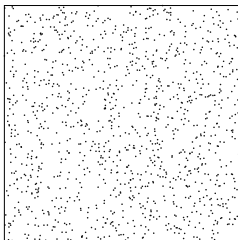
A RACS Ξ is called **stationary** if $\Xi \stackrel{d}{=} \Xi + x$, $\forall x \in \mathbb{R}^d$, and **isotropic** if $\Xi \stackrel{d}{=} g\Xi$, $\forall g \in SO(d)$

Theorem (Matheron (1975))

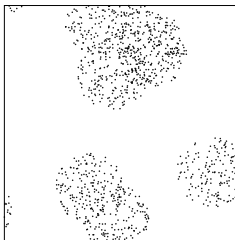
- ▶ *The RACS Ξ is stationary (isotropic) \iff
 $T_\Xi(C + x) = T_\Xi(C) \forall x \in \mathbb{R}^d$ and $T_\Xi(gC) = T_\Xi(C)$
 $\forall g \in SO(d)$, respectively*
- ▶ *Each stationary RACS $\Xi \neq \emptyset$ is a.s. unbounded*
- ▶ *For any stationary convex RACS Ξ , it holds $\Xi \in \{\emptyset, \mathbb{R}^d\}$ a.s.*

Examples

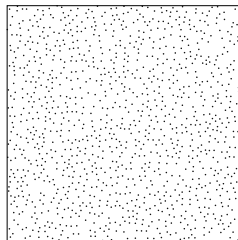
Stationary point processes in \mathbb{R}^2



Poisson process



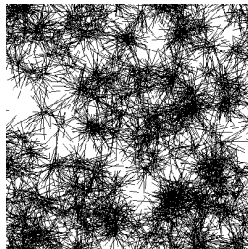
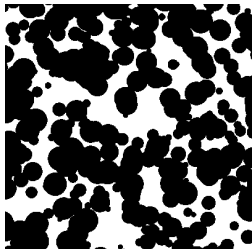
cluster process



hard-core process

Examples

Stationary germ–grain models in \mathbb{R}^2



Realizations of germ–grain models: Boolean model with spherical and polygonal grains, respectively; cluster process of segments

Examples

Germ–grain models

The germ–grain model $\Xi = \bigcup_{i=1}^{\infty} (\Xi_i + X_i)$ is called a **Boolean model** if

- ▶ the point process of germs $\{X_1, X_2, \dots\}$ is a stationary Poisson process in \mathbb{R}^d (with intensity λ)
- ▶ the grains Ξ_1, Ξ_2, \dots are i.i.d. and independent of $\{X_1, X_2, \dots\}$; $\Xi_i \stackrel{d}{=} \Xi_0$
- ▶ $\mathbb{E} |\Xi_0 \oplus K| < \infty$, $\forall K \in \mathcal{K}$.

Capacity functional: $T_{\Xi}(C) = 1 - e^{-\lambda \mathbb{E} |(-\Xi_0) \oplus C|}$, $\forall C \in \mathfrak{C}$

Specific intrinsic volumes

► Model assumptions

- Let Ξ be stationary, $\Xi \in \mathcal{S}$ a.s.
- $\mathbb{E} 2^{N(\Xi \cap [0,1]^d)} < \infty$, where $N(\emptyset) = 0$ and

$$N(K) = \min\{m \in \mathbb{N} : K = \bigcup_{i=1}^m K_i, K_i \in \mathcal{K}\} \text{ for } K \in \mathcal{R} \setminus \{\emptyset\}$$

- **Specific intrinsic volumes:** Let $\bar{V}_j(\Xi) = \lim_{n \rightarrow \infty} \frac{\mathbb{E} V_j(\Xi \cap W_n)}{|W_n|}$ for $j = 0, \dots, d$, where $\{W_n\}$ = sequence of monotonously increasing sampling windows $W_n = nW$ with $W \in \mathcal{K}$ and $|W| > 0$

In particular, $\bar{V}_d(\Xi) = P(o \in \Xi) = \mathbb{E}| \Xi \cap W | / |W|$

Estimation of $(\bar{V}_0(\Xi), \dots, \bar{V}_d(\Xi))^T$

Problem: Estimate $\bar{V}(\Xi) = (\bar{V}_0(\Xi), \dots, \bar{V}_d(\Xi))^T$ on the basis of a single sample from $\Xi \cap W$

Solution: For each $i = 0, \dots, d$, consider a random field $Y_i = \{Y_i(x), x \in \mathbb{R}^d\}$ such that

- ▶ Y_i is **stationary of second order**, i.e. $\mathbb{E} Y_i(x) = \mu_i$ and $\text{Cov}(Y_i(x), Y_i(x+h)) = \text{Cov}_{Y_i}(h) \quad \forall x, h \in \mathbb{R}^d$
- ▶ $\mu_i = \mathbb{E} Y_i(o) = \sum_{j=0}^d a_{ij} \bar{V}_j(\Xi)$, where the matrix $A = (a_{ij})_{i,j=0}^d$ is regular

Then, it holds $\bar{V}(\Xi) = A^{-1} \mu$, where $\mu = (\mu_0, \dots, \mu_d)^T$

Estimation of $(\bar{V}_0(\Xi), \dots, \bar{V}_d(\Xi))^T$

- ▶ For any $i = 0, \dots, d$ and $x \in U \subset W$, suppose that $Y_i(x)$ can be computed from $\Xi \cap W$
- ▶ Let $w(\cdot)$ be a probability measure with support in $U \subset W$

Examples: $w(\cdot) = |\cdot \cap U|/|U|$ and $w(\cdot) = \sum_{k=1}^m w_k \delta_{x_k}(\cdot)$ with $x_1, \dots, x_k \in U$, $w_1, \dots, w_m > 0$ and $w_1 + \dots + w_m = 1$

Then,

- ▶ $\hat{\mu} = (\hat{\mu}_0, \dots, \hat{\mu}_d)^T$ with $\hat{\mu}_i = \int_W Y_i(x) w(dx)$ is an unbiased estimator for $\mu = (\mu_0, \dots, \mu_d)^T$, and
- ▶ $\hat{V}(\Xi) = A^{-1} \hat{\mu}$ is unbiased for $\bar{V}(\Xi)$

Estimation of $(\bar{V}_0(\Xi), \dots, \bar{V}_d(\Xi))^T$

- ▶ **Method of least squares:** Consider $n > d$ random fields Y_i with the properties mentioned above
 - ▶ "Solve" the overdetermined system of linear equations
 - ▶ The solution is the LS-estimator for $(\bar{V}_0(\Xi), \dots, \bar{V}_d(\Xi))^T$
- ▶ **Minimization of variance**
 - ▶ Reduction of $Var(\hat{\mu}_i)$ by an appropriate choice of w
 - ▶ For a discrete averaging measure w : optimal weights w_1, \dots, w_m by **kriging of the mean** (Wackernagel (1998))
 - ▶ sampling points $x_1, \dots, x_m \in W$ by **optimal experimental design** for random fields (Näther (1985), Müller (2001))

Consistency

- ▶ Let $\mathbb{E} 4^{N(\Xi \cap [0,1]^d)} < \infty$ and $\int_{\mathbb{R}^d} |\text{Cov}_{Y_i}(h)| dh < \infty$
 $\forall i = 0, \dots, d$

Then,

- ▶ $\widehat{V}(\Xi, W_n) = A^{-1} \widehat{\mu}(W_n)$ with $W_n = nW$ and

$$\widehat{\mu}(W_n) = \left(\int_{W_n} Y_0(x) w(dx), \dots, \int_{W_n} Y_d(x) w(dx) \right)^T$$

is an L_2 -consistent estimator for $\overline{V}(\Xi)$, i.e.,

$$\mathbb{E} |\widehat{V}(\Xi, W_n) - \overline{V}(\Xi)|^2 \rightarrow 0, \quad n \rightarrow \infty$$

Asymptotic normality

If Ξ is a **stationary germ–grain model** with iid grains $\Xi_i \stackrel{d}{=} \Xi_0 \in \mathcal{K}$ that are independent of germs $\{X_1, X_2, \dots\}$ and some **additional assumptions** on the Y_i are fulfilled, then

$$\sqrt{|W_n|} (\widehat{V}(\Xi, W_n) - \overline{V}(\Xi)) \xrightarrow{d} N(0, A^{-1} \Sigma (A^{-1})^\top)$$

as $n \rightarrow \infty$ where

- ▶ $\Sigma = \left(\int_{\mathbb{R}^d} \text{Cov}_{Y_i Y_j}(h) dh \right)_{i,j=0}^d$ and
- ▶ $\text{Cov}_{Y_i Y_j}(h) = \mathbb{E} Y_i(x) Y_j(x+h) - \mu_i \mu_j$ is the **cross covariance function** of Y_i and Y_j

Asymptotic normality

Additional assumptions:

- ▶ $Y_j(x) = f_j((\Xi - x) \cap K_j)$, where the f_j are conditionally bounded valuations on \mathcal{R} ; $K_j \in \mathcal{K}$
 - ▶ $\mathbb{E} 2^{\rho N(\Xi \cap K_j)} < \infty$, where

$$N(\Xi \cap K_j) = \#\{i : (M_i + X_i) \cap K_j \neq \emptyset\}$$
 - ▶ $w(\cdot) = |\cdot \cap U_n| / |U_n|$, $U_n \subset W_n$
- and either
- ▶ $X = \{X_1, X_2, \dots\}$ is rapidly β -mixing
 - ▶ Ξ_0 uniformly bounded; $\rho = 2 + \delta$, $\delta > 0$
- or
- ▶ $X = \{X_1, X_2, \dots\}$ has finite range of correlation
 - ▶ $|\text{Cov}_{Y_i, Y_j}(h)| \leq g_{ij}(\Xi_0, h) \in L_1$ monotonously w.r.t. Ξ_0 ;
 $\rho = 2$

Consistent estimation of covariances

- ▶ Let $U_n \subset W_n$ be **averaging sets** such that $|U_n|^2/|W_n| \rightarrow 0$ and $\min_{h \in U_n} |W_n \cap (W_n - h)|/|W_n| \rightarrow 1$ for $n \rightarrow \infty$
- ▶ Then, the **L_2 -consistency** $\lim_{n \rightarrow \infty} \mathbb{E}|\widehat{\Sigma}_n - \Sigma|^2 = 0$ holds, where $\widehat{\Sigma}_n = (\widehat{\sigma}_{nij})_{i,j=0}^d$ with

$$\widehat{\sigma}_{nij} = \frac{1}{|W_n|} \int_{U_n} \widehat{\text{Cov}}_{nij}(h) |W_n \cap (W_n - h)| dh,$$

$$\begin{aligned} \widehat{\text{Cov}}_{nij}(h) &= \frac{\int_{W_n \cap (W_n + h)} Y_j(x) Y_i(x - h) dx}{|W_n \cap (W_n + h)|} \\ &= \frac{\int_{W_n} Y_i(x) dx \int_{W_n} Y_j(x) dx}{|W_n|^2} \end{aligned}$$

Example of random fields Y_i

Local Euler number

- ▶ Let $r_0, \dots, r_{d-1} > 0$ with $r_i \neq r_j, i \neq j$ and $|W \ominus B_{r_i}(o)| > 0$
- ▶ $Y_i(x) = V_0(\Xi \cap B_{r_i}(x))$ for $i = 0, \dots, d$
- ▶ **Edge-corrected estimator** (minus sampling):
 $U = W \ominus B_r(o)$

$$\hat{\mu}_i = \int_{W \ominus B_{r_i}(o)} V_0(\Xi \cap B_{r_i}(x)) w(dx) = \sum_{k=1}^m V_0(\Xi \cap B_{r_i}(x_k)) w_k$$

- ▶ Discrete averaging measure w , where
 $x_1, \dots, x_m \in W \ominus B_{r_i}(o)$ and, for example, $w_k = 1/m$,
 $k = 1, \dots, m$

Scan statistics for point processes

Scan statistic

Let $\Phi = \{X_i\}$ be an independently marked point process in \mathbb{R}^d with iid marks $\{M_i\}$ observed within a cube W . For a (cubic) subwindow $W_o \subset W$, define $S(W_o) = \sum_{i: X_i \in W_o} M_i$.

Scan statistic: $T = \sup_{W_o \in \mathcal{W}} S(W_o)$

- ▶ **Usual** scan statistic of fixed size $r > 0$:
 $\mathcal{W} = \{W_1 = x + r[0, 1]^d, x \in \mathbb{R}^d : W_1 \subset W\}$.
- ▶ **Multiscale** scan statistic: $\mathcal{W} = \{\text{all cubes } W_1 \subset W\}$

Limit theorems: $T = T_n \xrightarrow{d} ?$ as $W = W_n = n[0, 1]^d, n \rightarrow \infty$

Scan statistics for point processes

- ▶ **Scan statistics in \mathbb{R}^1 and \mathbb{R}^2** : Glaz, Balakrishnan (1999), Glaz, Naus, Wallenstein (2001)
- ▶ **LT for the usual scan statistic in \mathbb{R}^d** : $\Phi =$ stationary compound Poisson process (Chan (2007))
- ▶ **LT for the multiscale scan statistic in \mathbb{R}^1** : Cohen (1968), Iglehart (1972), Karlin, Dembo (1992), Doney, Maller (2005)
- ▶ **LT for the multiscale scan statistic in \mathbb{R}^d** : independently scattered Lévy measures (Kablichko, S. (2008))

Scan statistic for Lévy noise

- ▶ **Lévy noise:** Let $\xi = \{\xi(t), t \geq 0\}$ be a Lévy process with $\xi(0) = 0$, $\mathbb{E} \xi(1) = \mu$, $\sigma^2 = \text{Var} \xi(1) > 0$.
Lévy noise $\mathcal{Z} = \{\mathcal{Z}(B), B \in \mathcal{B}(\mathbb{R}^d)\}$ is an independently scattered stationary random measure on \mathbb{R}^d driven by ξ , i.e. $\mathcal{Z}(B) \stackrel{d}{=} \xi(|B|)$ for Borel sets $B \in \mathcal{B}(\mathbb{R}^d)$.
- ▶ **Multiscale scan statistic:**

$$T_n = \sup_{W_o \in \mathcal{W}_n} \mathcal{Z}(W_o), \quad n \in \mathbb{N}$$

for $\mathcal{W}_n = \{\text{all cubes within } W_n = [0, n]^d\}$.

LT for the scan statistic of Lévy noise

Theorem (Kabluchko, S. (2008))

- ▶ If $\mu > 0$ then $(T_n - \mu n^d)/(\sigma n^{d/2}) \xrightarrow{d} Y \sim N(0, 1)$
- ▶ If $\mu = 0$ then $T_n/(\sigma n^{d/2}) \xrightarrow{d} \sup_{W_o \in \mathcal{W}_1} Z(W_o)$, where $Z = \{Z(B), B \in \mathcal{B}(\mathbb{R}^d)\}$ is the standard Gaussian white noise on $[0, 1]^d$.
- ▶ If $\mu < 0$, the distribution of $\xi(1)$ is non-lattice, $\varphi(s) = \log \mathbb{E} e^{s\xi(1)}$ exists for $s \in [0, s_0)$ with the maximal $s_0 \in (0, \infty]$ and $\exists s^* \in (0, s_0): \varphi(s^*) = 0$ then

$$s^* T_n - d \log n - (d-1) \log \log n - c \xrightarrow{d} Y,$$

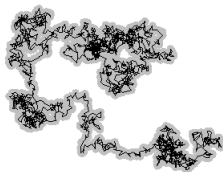
where Y is standard Gumbel distributed r. v. and c is a constant.

Open problems

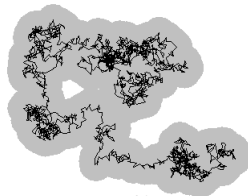
Limit theorems for the Wiener sausage

Let $S(T) = \{X(t) : t \in [0, T]\}$ be the path of the Brownian motion $X \subset \mathbb{R}^d$ with variance σ^2 up to time $T > 0$.

- ▶ **Wiener sausage** S_r of radius $r > 0$: $S_r = S(T) \oplus B_r(o)$



$r = 10$



$r = 40$

A realization of S_r

Intrinsic volumes of the Wiener sausage

- ▶ **Intrinsic volumes** $V_0(S_r), \dots, V_d(S_r)$ are well-defined a.s. for $d \leq 3, r > 0$;
 - ▶ $V_d(S_r) = |S_r|$
 - ▶ $2V_{d-1}(S_r) = \mathcal{H}^{d-1}(\partial S_r)$
 - ▶ $V_i(S_r) = (-1)^{d-i-1} V_i(\overline{\mathbb{R}^d \setminus S_r})$, $i = 0, \dots, d-2$, where ∂S_r is a Lipschitz manifold with $\text{reach}(\overline{\mathbb{R}^d \setminus S_r}) > 0$ a.s.
- ▶ **Compute** $\mathbb{E} V_i(S_r)$, $i = 0, \dots, d$.
 It is proved that $\mathbb{E} V_i(S_r) < \infty$, $i = d, d-1$ for all $d \geq 2$ and $\mathbb{E} V_0(S_r) < \infty$ for $d = 2$ (Rataj, Schmidt, Meschenmoser, S. (2005, 2009).

Mean volume of the Wiener sausage

- ▶ **Explicit formulae**
 - ▶ $d = 2$: Kolmogorov, Leontovich (1933)
 - ▶ $d = 3$: Spitzer (1964)
 - ▶ $d \geq 4$: Berezhkovskii *et al.* (1989)
- ▶ **Asymptotics of the volume**
 - ▶ Gettoor (1965)
 - ▶ Donsker, Varadhan (1975)
 - ▶ Le Gall (1988): **CLT for shrinking Wiener sausage** ($T \rightarrow \infty$ or $r \rightarrow 0$)
 - ▶ van den Berg, Bolthausen (1994)
 - ▶ ...

Other mean intrinsic volumes

- ▶ **Mean surface area:** Rataj, Schmidt, S. (2005)
- ▶ **Support measures and mean curvature functions:** Last (2005)
- ▶ **Mean intrinsic volumes $\mathbb{E} V_i(S_r)$ of lower order $i = 0, \dots, d - 2$: an open problem.**
Approximations can be obtained numerically (Rataj, Meschenmoser, S. (2009))

Mean surface area of the Wiener sausage

Theorem (Rataj, Schmidt, S. (2005))

For $d \geq 2$, it holds

$$2\mathbb{E}V_{d-1}(S_r) = d\kappa_d r^{d-1} + \frac{4d^2 \kappa_d r^{d-1}}{\pi^2} \int_0^\infty \frac{1 - e^{-\frac{\sigma^2 y^2 T}{2r^2}}}{y^3 (J_\nu^2(y) + Y_\nu^2(y))} dy$$

$$+ d\kappa_d \sigma^2 r^{d-3} T \left(\frac{(d-2)^2}{2} - \frac{4}{\pi^2} \int_0^\infty \frac{e^{-\frac{\sigma^2 y^2 T}{2r^2}}}{y (J_\nu^2(y) + Y_\nu^2(y))} dy \right)$$

for *almost all radii* $r > 0$, $\nu = (d - 2)/2$. For $d = 2, 3$, this formula holds for *all* $r > 0$. In the case $d = 3$, it simplifies to

$$\mathbb{E}V_2(S_r) = 4\pi r^2 + 8r\sigma\sqrt{2\pi T} + 2\pi\sigma^2 T.$$

Mean surface area of the Wiener sausage

- ▶ **Asymptotic behaviour** (Rataj, Schmidt, S. (2009))

$$2\mathbb{E}V_{d-1}(S_r) \sim \begin{cases} \pi\sigma^2 T r^{-1} \log^{-2} r & \text{if } d = 2, \\ 2\pi\sigma^2 T & \text{if } d = 3, \\ d\kappa_d\sigma^2 T \frac{(d-2)^2}{2} r^{d-3} & \text{if } d \geq 4 \end{cases} \quad \text{as } r \rightarrow 0.$$

- ▶ **LT for the volume** (Le Gall (1988)): for $d = 2$, it holds

$$(\log r)^2 (|S_r| + \pi / \log r) \xrightarrow{d} c - \pi^2 \gamma, \quad r \rightarrow 0,$$

where $\sigma^2 = T = 1$ and γ is the (renormalized) Brownian local time of self-intersections. For $d \geq 3$: CLT.

- ▶ **Open problem**: LT for the surface area $2V_{d-1}(S_r)$ and other intrinsic volumes $V_j(S_r)$ of the shrinking Wiener sausage!

Open problems

Limit theorems for excursion sets of stationary random fields

Let $X = \{X(t), t \in \mathbb{R}^d\}$ be a stationary C^2 -smooth random field, $A(X, u) = \{t \in \mathbb{R}^d : X(t) > u\}$, $u \in \mathbb{R}$ its excursion sets.

- ▶ $\mathbb{E} V_j(A(X, u))$ for Gaussian and related random fields: Adler, Taylor (2007)
- ▶ LT for $V_j(A(X, u))$:
 - ▶ Volume $V_d(A(X, u))$:
 - ▶ classical results for random processes and Gaussian random fields
 - ▶ (BL, θ) -dependent random fields: Bulinski, S., Timmermann; Meschenmoser, Shashkin (2009)
 - ▶ Other random fields: Open problem
 - ▶ Other intrinsic volumes: Open problem

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