

# The expected Euler characteristic approximation for Gaussian vector fields

Dan Cheng

Texas Tech University

(joint work with Yimin Xiao)

April 1, 2017

# 1. Introduction and Motivation

Let  $X = \{X(t), t \in T\}$  be a real-valued Gaussian random field, where  $T$  is the parameter set. For  $u > 0$ , how to evaluate the excursion probability

$$\mathbb{P} \left\{ \sup_{t \in T} X(t) \geq u \right\} = ?$$

- Double sum method [Pickands (1969), Piterbarg (1996), Chan and Lai (2006), etc.]
- Tube method [Sun (1993)]
- Rice method: the expected number of local maxima. See Piterbarg (1996), Adler (2000), Azaïs and Wschebor (2008), etc.
- Euler heuristic: the expected Euler characteristic of the excursion set. See Adler (1981, 2000), Worsley (1995), Taylor, Takemura and Adler (2005), Adler and Taylor (2007), etc.

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- Let  $A_u = \{t \in T : X(t) \geq u\}$  be the excursion set.
- Let  $\chi(A_u)$  and  $\mathbb{E}\{\chi(A_u)\}$  be the Euler characteristic and mean Euler characteristic of  $A_u$  respectively.
- In one dimension, the Euler characteristic is the number of connected components; and in two dimensions, it is the number of connected components minus the number of holes.

## Theorem [Taylor, Takemura and Adler (2005)]

Let  $X = \{X(t), t \in T\}$  be a centered smooth Gaussian random field with unit variance. There exists  $\alpha > 0$  such that

$$\mathbb{P} \left\{ \sup_{t \in T} X(t) \geq u \right\} = \mathbb{E}\{\chi(A_u)\} + o(e^{-\alpha u^2 - u^2/2}), \text{ as } u \rightarrow \infty.$$

- $\mathbb{E}\{\chi(A_u)\}$  is computable [Adler and Taylor (2007)],

$$\mathbb{E}\{\chi(A_u)\} = C_0 \Psi(u) + \sum_{j=1}^{\dim(T)} C_j u^{j-1} e^{-u^2/2},$$

where  $C_j$  are constants depending on  $X$  and  $T$ .

- When  $X$  is isotropic and  $T = [0, L]^N$ ,

$$\mathbb{E}\{\chi(A_u)\} = \Psi(u) + \sum_{j=1}^N \frac{\binom{N}{j} L^j \lambda^{j/2}}{(2\pi)^{(j+1)/2}} H_{j-1}(u) e^{-u^2/2},$$

where  $\lambda = \text{Var} \left( \frac{\partial X(t)}{\partial t_i} \right)$  and  $H_{j-1}(u)$  are Hermite polynomials.



Let  $\{(X(t), Y(s)) : t \in T, s \in S\}$  be an  $\mathbb{R}^2$ -valued, centered, unit-variance **smooth** Gaussian vector field, where  $T$  and  $S$  are rectangles in  $\mathbb{R}^N$ . For  $u > 0$ , how to evaluate the excursion probability

$$\mathbb{P} \left\{ \sup_{t \in T} X(t) \geq u, \sup_{s \in S} Y(s) \geq u \right\} = ?$$

## 2. Euler characteristic

- Consider rectangles  $T = \prod_{i=1}^N [a_i, b_i]$  and  $S = \prod_{i=1}^N [a'_i, b'_i]$ .

### Definition

A face  $J$  of dimension  $k$ , is defined by fixing a subset  $\sigma(J) \subset \{1, \dots, N\}$  of size  $k$  and a subset  $\varepsilon(J) = \{\varepsilon_j, j \notin \sigma(J)\} \subset \{0, 1\}^{N-k}$  of size  $N - k$ , so that

$$J = \{t \in T : a_j < t_j < b_j \text{ if } j \in \sigma(J), \\ t_j = (1 - \varepsilon_j)a_j + \varepsilon_j b_j \text{ if } j \notin \sigma(J)\}.$$

- Let  $\partial_k T$  be the collection of faces of dimension  $k$  in  $T$ , then  $\overset{\circ}{T} = \partial_N T$ ,  $\partial T = \bigcup_{k=0}^{N-1} \bigcup_{J \in \partial_k T} J$  and  $T = \bigcup_{k=0}^N \partial_k T = \bigcup_{k=0}^N \bigcup_{J \in \partial_k T} J$ .
- Similarly, let  $\partial_l S$  be the collection of faces of dimension  $l$  in  $S$ , then  $\overset{\circ}{S} = \partial_N S$ ,  $\partial S = \bigcup_{l=0}^{N-1} \bigcup_{L \in \partial_l S} L$  and  $S = \bigcup_{l=0}^N \partial_l S = \bigcup_{l=0}^N \bigcup_{L \in \partial_l S} L$ .

For a function  $f(\cdot) \in C^2(\mathbb{R}^N)$  and  $t \in \mathbb{R}^N$ , let

$$f_i(t) = \frac{\partial f(t)}{\partial t_i}, \quad f_{ij}(t) = \frac{\partial^2 f(t)}{\partial t_i \partial t_j}, \quad \forall i, j = 1, \dots, N,$$
$$\nabla f(t) = (f_1(t), \dots, f_N(t))^T, \quad \nabla^2 f(t) = (f_{ij}(t))_{i,j=1, \dots, N}.$$

For each  $t \in J \in \partial_k T$  and  $s \in L \in \partial_l S$ , let

$$\nabla X|_J(t) = (X_{i_1}(t), \dots, X_{i_k}(t))_{i_1, \dots, i_k \in \sigma(J)}^T, \quad \nabla^2 X|_J(t) = (X_{mn}(t))_{m, n \in \sigma(J)},$$
$$\nabla Y|_L(s) = (Y_{i_1}(s), \dots, Y_{i_l}(s))_{i_1, \dots, i_l \in \sigma(L)}^T, \quad \nabla^2 Y|_L(s) = (Y_{mn}(s))_{m, n \in \sigma(L)}.$$

Define the excursion sets

$$A_u(X, T) = \{t \in T : X(t) \geq u\},$$

$$A_u(Y, S) = \{s \in S : Y(s) \geq u\},$$

$$A_u(X, T) \times A_u(Y, S) = \{(t, s) \in T \times S : X(t) \geq u, Y(s) \geq u\}.$$

For each  $J \in \partial_k T$  and  $L \in \partial_l S$ , define the **number of extended outward critical point of index  $i$**  above level  $u$  as

$$\begin{aligned} \mu_i(X, J) := \#\{t \in J : X(t) \geq u, \nabla X|_J(t) = 0, \text{index}(\nabla^2 X|_J(t)) = i, \\ \varepsilon_j^* X_j(t) \geq 0 \text{ for all } j \notin \sigma(J)\}, \end{aligned}$$

$$\begin{aligned} \mu_i(Y, L) := \#\{s \in L : Y(s) \geq u, \nabla Y|_L(s) = 0, \text{index}(\nabla^2 Y|_L(s)) = i, \\ \varepsilon_j^* Y_j(s) \geq 0 \text{ for all } j \notin \sigma(L)\}, \end{aligned}$$

where where  $\varepsilon_j^* = 2\varepsilon_j - 1$  and the index of a matrix is defined as the number of its negative eigenvalues.

It follows from the Morse theorem [see Adler and Taylor (2007)] that the Euler characteristic of the excursion set can be represented as

$$\chi(A_u(X, T)) = \sum_{k=0}^N \sum_{J \in \partial_k T} (-1)^k \sum_{i=0}^k (-1)^i \mu_i(X, J),$$

$$\chi(A_u(Y, S)) = \sum_{l=0}^N \sum_{L \in \partial_l S} (-1)^l \sum_{i=0}^l (-1)^i \mu_i(Y, L).$$

Moreover,

$$\begin{aligned} \chi(A_u(X, T) \times A_u(Y, S)) &= \chi(A_u(X, T)) \times \chi(A_u(Y, S)) \\ &= \sum_{k,l=0}^N \sum_{J \in \partial_k T, L \in \partial_l S} (-1)^{k+l} \left( \sum_{i=0}^k (-1)^i \mu_i(X, J) \right) \left( \sum_{j=0}^l (-1)^j \mu_j(Y, L) \right). \end{aligned}$$

Therefore, we have

$$\begin{aligned} & \mathbb{E}\{\chi(A_u(X, T) \times A_u(Y, S))\} \\ &= \sum_{k,l=0}^N \sum_{J \in \partial_k T, L \in \partial_l S} (-1)^{k+l} \mathbb{E} \left\{ \sum_{i=0}^k \sum_{j=0}^l (-1)^{i+j} \mu_i(X, J) \mu_j(Y, L) \right\} \end{aligned}$$

### 3. Excursion probability

For each  $J \in \partial_k T$  and  $L \in \partial_l S$ , define the **number of extended outward maxima** above level  $u$  as

$$M_u^E(X, J) := \#\{t \in J : X(t) \geq u, \nabla X|_J(t) = 0, \text{index}(\nabla^2 X|_J(t)) = k, \\ \varepsilon_j^* X_j(t) \geq 0 \text{ for all } j \notin \sigma(J)\},$$
$$M_u^E(Y, L) := \#\{s \in L : Y(s) \geq u, \nabla Y|_L(s) = 0, \text{index}(\nabla^2 Y|_L(s)) = l, \\ \varepsilon_j^* Y_j(s) \geq 0 \text{ for all } j \notin \sigma(L)\}.$$

**Example:** Let  $T = [0, 1] = \{0\} \cup \{1\} \cup (0, 1)$ , then with probability 1,

$$\left\{ \sup_{t \in T} X(t) \geq u \right\} = \{X(0) \geq u, X'(0) \leq 0\} \cup \{X(1) \geq u, X'(1) \geq 0\} \\ \cup \{\exists t \in (0, 1) \text{ s.t. } X(t) \geq u, X'(t) = 0, X''(t) < 0\}.$$

- For Gaussian fields, since  $T = \bigcup_{k=0}^N \bigcup_{J \in \partial_k T} J$ , one can show that

$$\left\{ \sup_{t \in T} X(t) \geq u \right\} = \bigcup_{k=0}^N \bigcup_{J \in \partial_k T} \{M_u^E(J) \geq 1\} \quad \text{a.s.}$$

- For Gaussian vector fields, since  $T = \bigcup_{k=0}^N \bigcup_{J \in \partial_k T} J$  and  $S = \bigcup_{l=0}^N \bigcup_{L \in \partial_l S} L$ , one can show that

$$\left\{ \sup_{t \in T} X(t) \geq u, \sup_{s \in S} Y(s) \geq u \right\} \\ = \bigcup_{k,l=0}^N \bigcup_{J \in \partial_k T, L \in \partial_l S} \{M_u^E(X, J) \geq 1, M_u^E(Y, L) \geq 1\} \quad \text{a.s.}$$



Therefore, we obtain **the upper bound** for the excursion probability

$$\begin{aligned} & \mathbb{P} \left\{ \sup_{t \in T} X(t) \geq u, \sup_{s \in S} Y(s) \geq u \right\} \\ & \leq \sum_{k,l=0}^N \sum_{J \in \partial_k T, L \in \partial_l S} \mathbb{P} \{ M_u^E(X, J) \geq 1, M_u^E(Y, L) \geq 1 \} \\ & \leq \sum_{k,l=0}^N \sum_{J \in \partial_k T, L \in \partial_l S} \mathbb{E} \{ M_u^E(X, J) M_u^E(Y, L) \}. \end{aligned}$$

On the other hand, notice that

$$\begin{aligned}
 & \mathbb{E}\{M_u^E(X, J)M_u^E(Y, L)\} - \mathbb{P}\{M_u^E(X, J) \geq 1, M_u^E(Y, L) \geq 1\} \\
 &= \sum_{i,j=1}^{\infty} (ij - 1)\mathbb{P}\{M_u^E(X, J) = i, M_u^E(Y, L) = j\} \\
 &\leq \sum_{i,j=1}^{\infty} [i(i-1)j + j(j-1)i]\mathbb{P}\{M_u^E(X, J) = i, M_u^E(Y, L) = j\} \\
 &= \mathbb{E}\{M_u^E(X, J)[M_u^E(X, J) - 1]M_u^E(Y, L)\} \\
 &\quad + \mathbb{E}\{M_u^E(Y, L)[M_u^E(Y, L) - 1]M_u^E(X, J)\}
 \end{aligned} \tag{1}$$

and

$$\begin{aligned}
 & \mathbb{P}\{M_u^E(X, J) \geq 1, M_u^E(Y, L) \geq 1, M_u^E(X, J') \geq 1, M_u^E(Y, L') \geq 1\} \\
 &\leq \mathbb{P}\{M_u^E(X, J) \geq 1, M_u^E(Y, L) \geq 1, M_u^E(Y, L') \geq 1\} \\
 &\leq \mathbb{E}\{M_u^E(X, J)M_u^E(Y, L)M_u^E(Y, L')\}.
 \end{aligned} \tag{2}$$

By the Bonferroni inequality, (1) and (2), we obtain **the lower bound** for the excursion probability

$$\begin{aligned}
 & \mathbb{P} \left\{ \sup_{t \in T} X(t) \geq u, \sup_{s \in S} Y(s) \geq u \right\} \\
 & \geq \sum_{k,l=0}^N \sum_{J \in \partial_k T, L \in \partial_l S} \left\{ \mathbb{E} \{ M_u^E(X, J) M_u^E(Y, L) \} - \mathbb{E} \{ M_u^E(X, J) \right. \\
 & \quad \times [M_u^E(X, J) - 1] M_u^E(Y, L) \} - \mathbb{E} \{ M_u^E(Y, L) [M_u^E(Y, L) - 1] M_u^E(X, J) \} \left. \right\} \\
 & \quad - \sum_{k,k',l=0}^N \sum_{\substack{J \in \partial_k T, L \in \partial_l S \\ J' \in \partial_{k'} T, J \neq J'}} \mathbb{E} \{ M_u^E(X, J) M_u^E(X, J') M_u^E(Y, L) \} \\
 & \quad - C_N \sum_{k,l,l'=0}^N \sum_{\substack{J \in \partial_k T, L \in \partial_l S \\ L' \in \partial_{l'} S, L \neq L'}} \mathbb{E} \{ M_u^E(X, J) M_u^E(Y, L) M_u^E(Y, L') \},
 \end{aligned}$$

where  $C_N$  is a constant depending only on  $N$ .

Let

$$r(t, s) = \mathbb{E}\{X(t)Y(s)\}, \quad \rho = \sup_{t \in T, s \in S} \mathbb{E}\{X(t)Y(s)\}.$$

- We call a function  $h(u)$  *super-exponentially small* [when compared with the excursion probability  $\mathbb{P}\{\sup_{t \in T} X(t) \geq u, \sup_{s \in S} Y(s) \geq u\}$ ], if there exists a constant  $\alpha > 0$  such that  $h(u) = o(e^{-\alpha u^2 - u^2/(1+\rho)})$  as  $u \rightarrow \infty$ .
- The sketch of proof consists of the following two steps: (i) all terms, except for those in the upper bound, are super-exponentially small; (ii) the difference between the upper bound and the expected Euler characteristic of the excursion set is also super-exponentially small.

**Lemma** Under certain smoothness and regularity conditions, there exists some  $\alpha > 0$  such that

$$\begin{aligned} & \mathbb{P} \left\{ \sup_{t \in T} X(t) \geq u, \sup_{s \in S} Y(s) \geq u \right\} \\ &= \sum_{k,l=0}^N \sum_{J \in \partial_k T, L \in \partial_l S} \mathbb{E} \{ M_u^E(X, J) M_u^E(Y, L) \} + o \left( \exp \left\{ -\frac{u^2}{1 + \rho} - \alpha u^2 \right\} \right) \end{aligned}$$

and

$$\begin{aligned} & \sum_{k,l=0}^N \sum_{J \in \partial_k T, L \in \partial_l S} \mathbb{E} \{ M_u^E(X, J) M_u^E(Y, L) \} \\ &= \mathbb{E} \{ \chi(A_u(X, T) \times A_u(Y, S)) \} + o \left( \exp \left\{ -\frac{u^2}{1 + \rho} - \alpha u^2 \right\} \right). \end{aligned}$$

**Theorem** Let  $\{(X(t), Y(s)) : t \in T, s \in S\}$  be an  $\mathbb{R}^2$ -valued, centered, unit-variance Gaussian vector field. Then under certain smoothness and regularity conditions, there exists  $\alpha > 0$  such that

$$\begin{aligned} & \mathbb{P} \left\{ \sup_{t \in T} X(t) \geq u, \sup_{s \in S} Y(s) \geq u \right\} \\ &= \mathbb{E} \{ \chi(A_u(X, T) \times A_u(Y, S)) \} + o \left( \exp \left\{ -\frac{u^2}{1 + \rho} - \alpha u^2 \right\} \right). \end{aligned}$$

- The expected Euler characteristic approximation  $\mathbb{E}\{\chi(A_u(X, T) \times A_u(Y, S))\}$  can be written as an integral by the Kac-Rice formula. However, due to the correlation function between  $X$  and  $Y$ , it is usually difficult to obtain the exact value. We usually need to apply the Laplace method to approximate the integral.
- The expected Euler characteristic approximation can be extended to general  $\mathbb{R}^d$ -valued Gaussian vector fields.
- Extensions for more general  $T$  and  $S$  are possible (such as spheres).

Thank you!