

On the Distribution of the Maximum of a Gaussian Field with d Parameters.*

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Abstract

Let I be a compact d -dimensional manifold, $X : I \rightarrow \mathcal{R}$ a Gaussian process with regular paths and $F_I(u)$, $u \in \mathcal{R}$ the probability distribution function of $\sup_{t \in I} X(t)$.

We prove that under certain regularity and non-degeneracy conditions, F_I is a C^1 -function and satisfies a certain implicit equation that permits to give bounds for its values and to compute its asymptotic behaviour as $u \rightarrow +\infty$. This is a partial extension of previous results by the authors in the case $d = 1$.

Our methods use strongly the so-called Rice formulae for the moments of the number of roots of an equation of the form $Z(t) = x$, where $Z : I \rightarrow \mathcal{R}^d$ is a random field and x a fixed point in \mathcal{R}^d . We also give proofs for this kind of formulae, which have their own interest beyond the present application.

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1 Introduction and notations.

Let I be a d -dimensional compact manifold and $X : I \rightarrow \mathcal{R}$ a Gaussian process with regular paths defined on some probability space $(\Omega, \mathcal{A}, \mathbb{P})$. Define $M_I = \sup_{t \in I} X(t)$ and $F_I(u) = \mathbb{P}\{M_I \leq u\}$, $u \in \mathcal{R}$ the probability distribution function of the random variable M_I . Our aim is to study the regularity of the function F_I when $d > 1$.

There exist a certain number of general results on this subject, starting from the papers by Ylvisaker (1968) and Tsirelson (1975) (see also Weber (1985), Lifshits (1995), Diebolt and Posse (1996) and references therein). The main purpose of this paper is to extend to $d > 1$ some of the results about the regularity of the function $u \rightsquigarrow F_I(u)$ in Azaïs & Wschebor (2001), which concern the case $d = 1$.

Our main tool here is Rice Formula for the moments of the number of roots $N_u^Z(I)$ of the equation $Z(t) = u$ on the set I , where $\{Z(t) : t \in I\}$ is an \mathcal{R}^d -valued Gaussian field, I is a subset of \mathcal{R}^d and u a given point in \mathcal{R}^d . For $d > 1$, even though it has been used in various contexts, as far as the authors know, a full proof of Rice Formula for the moments of $N_u^Z(I)$ seems to have only been published by R. Adler (1981) for the first moment of the number of critical points of a real-valued stationary Gaussian process with a d -dimensional parameter, and extended by Azaïs and Delmas (2002) to the case of processes with constant variance. Cabaña (1985) contains related formulae for random fields; see also the PHD thesis of Konakov cited by Piterbarg (1996b). In the next section we give a more general result which has an interest that goes beyond the application of the present paper. At the same time the proof appears to be simpler than previous ones. We have also included the proof of the formula for higher moments, which in fact follows easily from the first moment. Both extend with no difficulties to certain classes of non-Gaussian processes.

It should be pointed out that the validity of Rice Formula for Lebesgue-almost every $u \in \mathcal{R}^d$ is easy to prove (Brillinger, 1972) but this is insufficient for a certain number of standard applications. For example, assume $X : I \rightsquigarrow \mathcal{R}$ is a real-valued random process and one is willing to compute the moments of the number of critical points of X . Then, we must take for Z the random field $Z(t) = X'(t)$ and the formula one needs is for the precise value $u = 0$ so that a formula for almost every u does not solve the problem.

We have added Rice Formula for processes defined on smooth manifolds. Even though Rice Formula is local, this is convenient for various applications. We will need a formula of this sort to state and prove the implicit formulae for the derivatives of the distribution of the maximum (see Section 3).

The results on the differentiation of F_I are partial extensions of Azaïs & Wschebor (2001). Here, we have only considered the first derivative $F'_I(u)$. In fact one can push our procedure one step more and prove the existence of $F''_I(u)$ as well as some implicit formula for it. But we have not included this in the present paper since formulae become very complicated and it is unclear at present whether the actual computations can be performed, even in simple examples. The technical reason for this is that, following the present method, to compute $F''_I(u)$, one needs to differentiate expressions that contain the "helix process" that we introduce in section 4, containing singularities with unpleasant behaviour, (see Azaïs and Wschebor, 2002).

For Gaussian fields defined on a d -dimensional regular manifold ($d > 1$) and possessing regular paths we obtain some improvements with respect to classical and general results due to Tsirelson (1975) for Gaussian sequences. An example is Corollary 5.1, that provides an asymptotic formula for $F'_I(u)$ as $u \rightarrow +\infty$ which is explicit in terms of the covariance of the process and can be compared with Theorem 4 in Tsirelson (1975) where an implicit expression depending on the function F itself is given.

We use the following notations:

If Z is a smooth function $U \rightsquigarrow \mathcal{R}^d$, U a subset of \mathcal{R}^d , its successive derivatives are denoted $Z', Z'', \dots, Z^{(k)}$ and considered respectively as linear, bilinear, ..., k -linear forms on \mathcal{R}^d . For example, $X^{(3)}(t)[v_1, v_2, v_3]$ is the value of the third derivative at point t applied to the triplet (v_1, v_2, v_3) . The same notation is used for a derivative on a \mathcal{C}^∞ manifold.

\dot{I} , ∂I and \bar{I} are respectively the interior, the boundary and the closure of the set I . If ξ is a random vector with values in \mathcal{R}^d , whenever they exist, we denote by $p_\xi(x)$ the value of the density of ξ at the point x , by $E(\xi)$ its expectation and by $\text{Var}(\xi)$ its variance-covariance matrix. λ is Lebesgue measure. If u, v are points in \mathcal{R}^d , $\langle u, v \rangle$ denotes their usual scalar product and $\|u\|$ the Euclidean norm of u . For M a $d \times d$ real matrix, we denote $\|M\| = \sup_{\|x\|=1} \|Mx\|$

Also for symmetric M , $M \succ 0$ (respectively $M \prec 0$) denotes that M is positive definite (resp. negative definite). A^c denotes the complement of the set A . For real x , $x^+ = \sup(x, 0)$, $x^- = \sup(-x, 0)$

2 Rice formulae

Our main results in this section are the following:

Theorem 2.1 Let $Z : I \rightsquigarrow \mathcal{R}^d$, I a compact subset of \mathcal{R}^d , be a random field and $u \in \mathcal{R}^d$. Assume that:

A0: Z is Gaussian, A1: $t \rightsquigarrow Z(t)$ is a.s. of class \mathcal{C}^1 ,

A2: for each $t \in I$, $Z(t)$ has a non degenerate distribution (i.e. $\text{Var}(Z(t)) \succ 0$),

A3: $\text{P}\{\exists t \in I, Z(t) = u, \det(Z'(t)) = 0\} = 0$

A4: $\lambda(\partial I) = 0$.

Then

$$\mathbb{E}(N_u^Z(I)) = \int_I \mathbb{E}(|\det(Z'(t))| | Z(t) = u) p_{Z(t)}(u) dt, \quad (1)$$

and both members are finite.

Theorem 2.2 Let $k, k \geq 2$ be an integer. Assume the same hypotheses as in Theorem (2.1) excepting for A2 that is replaced by

A'2 : for $t_1, \dots, t_k \in I$ pairwise different values of the parameter, the distribution of $(Z(t_1), \dots, Z(t_k))$ does not degenerate in $(\mathcal{R}^d)^k$. Then

$$\begin{aligned} & \mathbb{E}[(N_u^Z(I))(N_u^Z(I) - 1) \dots (N_u^Z(I) - k + 1)] \\ &= \int_{I^k} \mathbb{E} \left(\prod_{j=1}^k |\det(Z'(t_j))| | Z(t_1) = \dots = Z(t_k) = u \right) \\ & \quad p_{Z(t_1), \dots, Z(t_k)}(u, \dots, u) dt_1 \dots dt_k, \quad (2) \end{aligned}$$

where both members may be infinite.

Remark.

Note that Theorem 2.1 (resp 2.2) remains valid, excepting for the finiteness of the expectation in Theorem (2.1), if I is open and hypotheses A0,A1,A2 (resp A'2) and A3 are verified. This follows immediately from the above statements. A standard extension argument shows that (1) holds true if one replaces I by any Borel subset of \dot{I}

Sufficient conditions for hypotheses A3 to hold are given by the next proposition. Under condition a) the result is proved in Lemma 5 of Cucker and Wschebor (2002). Under condition b) the proof is straightforward.

Proposition 2.1 Let $Z : I \rightsquigarrow \mathcal{R}^d$, I a compact subset of \mathcal{R}^d be a random field with paths of class \mathcal{C}^1 and $u \in \mathcal{R}^d$. Assume that

- $p_{Z(t)}(x) \leq C$ for all $t \in I$ and x in some neighbourhood of u .

- at least one of the two following hypotheses is satisfied:

a) a.s. $t \rightsquigarrow Z(t)$ is of class \mathcal{C}^2

b) $\alpha(\delta) = \sup_{t \in I, x \in V(u)} \mathbb{P}\{|\det(Z'(t))| < \delta/Z(t) = x\} \rightarrow 0$ as $\delta \rightarrow 0$, where $V(u)$ is some neighbourhood of u .

Then A3 holds true.

The following lemma is easy to prove:

Lemma 2.1 *With the notations of Theorem (2.1), suppose that A1 and A4 hold true and that $p_{Z(t)}(x) \leq C$ for all $t \in I$ and x in some neighbourhood of u . Then $\mathbb{P}\{N_u^Z(\partial I) \neq 0\} = 0$*

Lemma 2.2 *Let $Z : I \rightarrow \mathcal{R}^d$, I a compact subset of \mathcal{R}^d , be a \mathcal{C}^1 function and u a point in \mathcal{R}^d . Assume that*

a) $\inf_{t \in Z^{-1}(\{u\})} \left(\lambda_{\min}(Z'(t)) \right) \geq \Delta > 0$

b) $\omega_{Z'}(\eta) < \Delta/d$

where $\omega_{Z'}$ is the continuity modulus of Z' , defined as the maximum of the continuity moduli of its entries, $\lambda_{\min}(M)$ is the square root of smallest eigenvalue of $M^T M$ and η is a positive number.

Then, if t_1, t_2 are two distinct roots of the equation $Z(t) = u$ such that the segment $[t_1, t_2]$ is contained in I , the Euclidean distance between t_1 and t_2 is greater than η .

Proof: Set $\tilde{\eta} = \|t_1 - t_2\|$, $v = \frac{t_1 - t_2}{\|t_1 - t_2\|}$. Using the mean value theorem, for $i = 1, \dots, d$, there exists $\xi_i \in [t_1, t_2]$ such that $(Z'(\xi_i)v)_i = 0$. Thus

$$\begin{aligned} |(Z'(t_1)v)_i| &= |(Z'(t_1)v)_i - (Z'(\xi_i)v)_i| \\ &\leq \sum_{k=1}^d |Z'(t_1)_{ik} - Z'(\xi_i)_{ik}| |v_k| \leq \omega_{Z'}(\tilde{\eta}) \sum_{k=1}^d |v_k| \leq \omega_{Z'}(\tilde{\eta}) \sqrt{d} \end{aligned}$$

In conclusion $\Delta \leq \lambda_{\min}(Z'(t_1)) \leq \|Z'(t_1)v\| \leq \omega_{Z'}(\tilde{\eta})d$, that implies $\tilde{\eta} > \eta$. \square

Proof of Theorem 2.1: Consider a continuous non-decreasing function F such that $F(x) = 0$ for $x \leq 1/2$ $F(x) = 1$ for $x \geq 1$. Let Δ and η be positive real numbers. Define the random function

$$\alpha_{\Delta, \eta}(u) = F\left(\frac{1}{2\Delta} \inf_{s \in I} [\lambda_{\min}(Z'(s)) + \|Z(s) - u\|]\right) \times \left(1 - F\left(\frac{d}{\Delta} \omega_{Z'}(\eta)\right)\right), \quad (3)$$

and the set $I_{-\eta} = \{t \in I : \|t - s\| \geq \eta, \forall s \notin I\}$. If $\alpha_{\Delta,\eta}(u) > 0$ and $N_u^Z(I_{-\eta})$ does not vanish, conditions a) and b) in Lemma 2.2 are satisfied. Hence, in each ball with diameter $\frac{\eta}{2}$ centred at a point in $I_{-\eta}$ there is at most one root of the equation $Z(t) = u$, and a compactness argument shows that $N_u^Z(I_{-\eta})$ is bounded by a constant $C(\eta, I)$, depending only on η and on the set I .

Take now any real-valued non-random continuous function $f : \mathcal{R}^d \rightarrow \mathcal{R}$ with compact support. Because of the coarea formula (Federer, 1969, Th 3.2.3), since a.s. Z is Lipschitz and $\alpha_{\Delta,\eta}(u) \cdot f(u)$ is integrable:

$$\int_{\mathcal{R}^d} f(u) N_u^Z(I_{-\eta}) \alpha_{\Delta,\eta}(u) du = \int_{I_{-\eta}} |\det(Z'(t))| f(Z(t)) \alpha_{\Delta,\eta}(Z(t)) dt.$$

Taking expectations in both sides,

$$\begin{aligned} \int_{\mathcal{R}^d} f(u) \mathbb{E} (N_u^Z(I_{-\eta}) \alpha_{\Delta,\eta}(u)) du &= \\ &= \int_{\mathcal{R}^d} f(u) du \int_{I_{-\eta}} \mathbb{E} (|\det(Z'(t))| \alpha_{\Delta,\eta}(u) / Z(t) = u) p_{Z(t)}(u) dt. \end{aligned}$$

It follows that the two functions

$$(i) : \mathbb{E} (N_u^Z(I_{-\eta}) \alpha_{\Delta,\eta}(u))$$

$$(ii) : \int_{I_{-\eta}} \mathbb{E} (|\det(Z'(t))| \alpha_{\Delta,\eta}(u) / Z(t) = u) p_{Z(t)}(u) dt,$$

coincide Lebesgue-almost everywhere as functions of u .

Let us prove that both functions are continuous, hence they are equal for every $u \in \mathcal{R}^d$.

Fix $u = u_0$ and let us show that the function in (i) is continuous at $u = u_0$. Consider the random variable inside the expectation sign in (i). Almost surely, there is no point t in $Z^{-1}(\{u_0\})$ such that $\det(Z'(t)) = 0$. By the local inversion theorem, $Z(\cdot)$ is invertible in some neighbourhood of each point belonging to $Z^{-1}(\{u_0\})$ and the distance from $Z(t)$ to u_0 is bounded below by a positive number for $t \in I_{-\eta}$ outside of the union of these neighbourhoods. This implies that, a.s., as a function of u , $N_u^Z(I_{-\eta})$ is constant in some (random) neighbourhood of u_0 . On the other hand, it is clear from its definition that the function $u \rightsquigarrow \alpha_{\Delta,\eta}(u)$ is continuous and bounded. We may now apply dominated convergence as $u \rightarrow u_0$, since $N_u^Z(I_{-\eta}) \alpha_{\Delta,\eta}(u)$ is bounded by a constant that does not depend on u .

For the continuity of (ii), it is enough to prove that, for each $t \in I$ the conditional expectation in the integrand is a continuous function of u . Note that the random variable $|\det(Z'(t))|\alpha_{\Delta,\eta}(u)$ is a functional defined on $\{(Z(s), Z'(s)) : s \in I\}$. Perform a Gaussian regression of $(Z(s), Z'(s)) : s \in I$ with respect to the random variable $Z(t)$, that is, write

$$\begin{aligned} Z(s) &= Y^t(s) + \alpha^t(s)Z(t) \\ Z'_j(s) &= Y_j^t(s) + \beta_j^t(s)Z(t), \quad j = 1, \dots, d \end{aligned}$$

where $Z'_j(s)$ ($j = 1, \dots, d$) denote the columns of $Z'(s)$, $Y^t(s)$ and $Y_j^t(s)$ are Gaussian vectors, independent of $Z(t)$ for each $s \in I$, and the regression matrices $\alpha^t(s)$, $\beta_j^t(s)$ ($j = 1, \dots, d$) are continuous functions of s, t (take into account A2). Replacing in the conditional expectation we are now able to get rid of the conditioning, and using the fact that the moments of the supremum of an a.s. bounded Gaussian process are finite, the continuity in u follows by dominated convergence.

So, now we fix $u \in \mathcal{R}^d$ and make $\eta \downarrow 0$, $\Delta \downarrow 0$ in that order, both in (i) and (ii). For (i) one can use Beppo Levi's Theorem. Note that almost surely $N_u^Z(I_{-\eta}) \uparrow N_u^Z(I) = N_u^Z(I)$, where the last equality follows from Lemma 2.1. On the other hand, the same Lemma 2.1 plus A3 imply together that, almost surely:

$$\inf_{s \in I} \left[\lambda_{\min}(Z'(s)) + \|Z(s) - u\| \right] > 0$$

so that the first factor in the right-hand member of (3) increases to 1 as Δ decreases to zero. Hence by Beppo Levi's Theorem:

$$\lim_{\Delta \downarrow 0} \lim_{\eta \downarrow 0} \mathbb{E} \left(N_u^Z(I_{-\eta}) \alpha_{\Delta,\eta}(u) \right) = \mathbb{E} \left(N_u^Z(I) \right).$$

For (ii), one can proceed in a similar way after de-conditioning obtaining (1). To finish the proof, remark that standard Gaussian calculations show the finiteness of the right-hand member of (1). \square

Proof of Theorem 2.2: For each $\delta > 0$, define the domain

$$D_{k,\delta}(I) = \{(t_1, \dots, t_k) \in I^k, \|t_i - t_j\| \geq \delta \text{ if } i \neq j, i, j = 1, \dots, k\}$$

and the process \tilde{Z}

$$(t_1, \dots, t_k) \in D_{k,\delta}(I) \rightsquigarrow \tilde{Z}(t_1, \dots, t_k) = (Z(t_1), \dots, Z(t_k)).$$

It is clear that \tilde{Z} satisfies the hypotheses of Theorem 2.1 for every value $(u, \dots, u) \in (\mathcal{R}^d)^k$. So,

$$\begin{aligned} \mathbb{E} \left[N_{(u, \dots, u)}^{\tilde{Z}}(D_{k, \delta}(I)) \right] &= \int_{D_{k, \delta}(I)} \\ \mathbb{E} \left(\prod_{j=1}^k |\det(Z'(t_j))| / Z(t_1) = \dots = Z(t_k) = u \right) &p_{Z(t_1), \dots, Z(t_k)}(u, \dots, u) dt_1 \dots dt_k \end{aligned} \quad (4)$$

To finish, let $\delta \downarrow 0$, note that $(N_u^Z(I))(N_u^Z(I) - 1) \dots (N_u^Z(I) - k + 1)$ is the monotone limit of $N_{(u, \dots, u)}^{\tilde{Z}}(D_{k, \delta}(I))$, and that the diagonal $D_k(I) = \{(t_1, \dots, t_k) \in I^k, t_i = t_j \text{ for some pair } i, j, i \neq j\}$ has zero Lebesgue measure in $(\mathcal{R}^d)^k$. \square

Remark

Even though we will not use this in the present paper, we point out that it is easy to adapt the proofs of Theorems 2.1 and 2.2 to certain classes of non-Gaussian processes.

For example, the statement of Theorem 2.1 remains valid if one replaces hypotheses A0 and A2 respectively by the following B0 and B2:

B0 : $Z(t) = H(Y(t))$ for $t \in I$ where

$Y : I \rightarrow \mathcal{R}^n$ is a Gaussian process with \mathcal{C}^1 paths such that for each $t \in I$, $Y(t)$ has a non-degenerate distribution and $H : \mathcal{R}^n \rightarrow \mathcal{R}^d$ is a \mathcal{C}^1 function.

B2 : for each $t \in I$, $Z(t)$ has a density $p_{Z(t)}$ which is continuous as a function of (t, u) .

Note that B0 and B2 together imply that $n \geq d$. The only change to be introduced in the proof of the theorem is in the continuity of (ii) where the regression is performed on $Y(t)$ instead of $Z(t)$

Similarly, the statement of Theorem 2.2 remains valid if we replace A0 by B0 and add the requirement the joint density of $Z(t_1), \dots, Z(t_k)$ to be a continuous function of t_1, \dots, t_k, u for pairwise different t_1, \dots, t_k

Now consider a process X from I to \mathcal{R} and define

$$M_{u,1}^X(I) = \# \{t \in I, X(\cdot) \text{ has a local maximum at the point } t, X(t) > u\}$$

$$M_{u,2}^X(I) = \# \{t \in I, X'(t) = 0, X(t) > u\}$$

The problem of writing Rice Formulae for the factorial moments of these random variables can be considered as a particular case of the previous one and the proofs are

the same, mutatis mutandis. For further use, we state as a theorem, Rice Formula for the expectation. For short we do not state the equivalent of Theorem (2.2) that holds true similarly.

Theorem 2.3 *Let $X : I \rightsquigarrow \mathcal{R}$, I a compact subset of \mathcal{R}^d , be a random field. Let $u \in \mathcal{R}$, define $M_{u,i}^X(I)$, $i = 1, 2$ as above. For each $d \times d$ real symmetric matrix M , we put $\delta^1(M) := |\det(M)| \mathbb{1}_{M < 0}$, $\delta^2(M) := |\det(M)|$.*

Assume:

A0: X is Gaussian, A"1: a.s. $t \rightsquigarrow X(t)$ is of class \mathcal{C}^2 ,

A"2: for each $t \in I$, $X(t), X'(t)$ has a non degenerate distribution in $\mathcal{R}^1 \times \mathcal{R}^d$,

A"3: either a.s. $t \rightsquigarrow X(t)$ is of class \mathcal{C}^3

or $\alpha(\delta) = \sup_{t \in I, x' \in V(0)} \mathbb{P}(|\det(X''(t))| < \delta / X'(t) = x') \rightarrow 0$ as $\delta \rightarrow 0$, where $V(0)$ denotes some neighbourhood of 0,

A4: ∂I has zero Lebesgue measure.

Then, for $i = 1, 2$:

$$\mathbb{E}(M_{u,i}^X(I)) = \int_u^\infty dx \int_I \mathbb{E}(\delta^i(X''(t)) / X(t) = x, X'(t) = 0) p_{X(t), X'(t)}(x, 0) dt$$

and both members are finite.

2.1 Processes defined on a smooth manifold.

Let U be a differentiable manifold (by differentiable we mean infinitely differentiable) of dimension d . We suppose that U is orientable in the sense that there exists a non-vanishing differentiable d -form Ω on U . This is equivalent to assuming that there exists an atlas $((U_i, \phi_i); i \in I)$ such that for any pair of intersecting charts $(U_i, \phi_i), (U_j, \phi_j)$, the Jacobian of the map $\phi_i \circ \phi_j^{-1}$ is positive.

We consider a Gaussian stochastic process with real values and \mathcal{C}^2 paths $X = \{X(t) : t \in U\}$ defined on the manifold U . In this subsection we first write Rice Formulae for this kind of processes without further hypotheses on U . When U is equipped with a Riemannian metric, we give without details and proof, a nicer form. Other forms exist also when U is naturally embedded in a Euclidean space, but we do not need this in the sequel. (see Azaïs and Wschebor, 2002).

We will assume that in every chart $X(t)$ and $DX(t)$ have a non-degenerate joint distribution and that hypothesis A"3 is verified. For S a Borel subset of \dot{U} , the following quantities are well defined and measurable: $M_{u,1}^X(S)$, the number of local maxima and $M_{u,2}^X(S)$, the number of critical points.

Proposition 2.2 For $k = 1, 2$ the quantity which is expressed in every chart ϕ with coordinates s_1, \dots, s_d as

$$\int_u^{+\infty} dx \mathbb{E} (\delta^k(Y''(s))/Y(s) = x, Y'(s) = 0) p_{Y(s), Y'(s)}(x, 0) \wedge_{i=1}^d ds_i, \quad (5)$$

where $Y(s)$ is the process X written in the chart : $Y = X \circ \phi^{-1}$, defines a d -form Ω^k on \dot{U} and for every Borel set $S \subset \dot{U}$

$$\int_S \Omega^k = \mathbb{E} (M_{u,k}^X(S)).$$

Proof: Note that a d -form is a measure on \dot{U} whose image in each chart is absolutely continuous with respect to Lebesgue measure $\wedge_{i=1}^d ds_i$. To prove that (5) defines an d -form, it is sufficient to prove that its density with respect to $\wedge_{i=1}^d ds_i$, satisfies locally the change-of-variable formula. Let $(U_1, \phi_1), (U_2, \phi_2)$ two intersecting charts and set

$$U_3 := U_1 \cap U_2 ; Y_1 := X \circ \phi_1^{-1} ; Y_2 := X \circ \phi_2^{-1} ; H := \phi_2 \circ \phi_1^{-1}.$$

Denote by s_i^1 and s_i^2 , $i = 1, \dots, d$ the coordinates in each chart. We have

$$\begin{aligned} \frac{\partial Y_1}{\partial s_i^1} &= \sum_{i'} \frac{\partial Y_2}{\partial s_{i'}^2} \frac{\partial H_{i'}}{\partial s_i^1} \\ \frac{\partial^2 Y_1}{\partial s_i^1 \partial s_j^1} &= \sum_{i', j'} \frac{\partial^2 Y_2}{\partial s_{i'}^2 \partial s_{j'}^2} \frac{\partial H_{i'}}{\partial s_i^1} \frac{\partial H_{j'}}{\partial s_j^1} + \sum_{i'} \frac{\partial Y_2}{\partial s_{i'}^2} \frac{\partial^2 H_{i'}}{\partial s_i^1 \partial s_j^1}. \end{aligned}$$

Thus at every point

$$\begin{aligned} Y_1'(s^1) &= (H'(s^1))^T Y_2'(s^2), \\ p_{Y_1(s^1), Y_1'(s^1)}(x, 0) &= p_{Y_2(s^2), Y_2'(s^2)}(x, 0) |\det(H'(s^1))|^{-1} \end{aligned}$$

and at a singular point

$$Y_1''(s^1) = (H'(s^1))^T Y_2''(s^2) H'(s^1).$$

On the other hand, by the change of variable formula

$$\wedge_{i=1}^d ds_i^1 = |\det(H'(s^1))|^{-1} \wedge_{i=1}^d ds_i^2.$$

Replacing in the integrand in (5), one checks the desired result.

For the second part again it suffices to prove it locally for an open subset S included in a unique chart. Let (S, ϕ) a chart and let again $Y(s)$ be the process written in this chart, it suffices to check that

$$\mathbb{E} (M_{u,k}^X(S)) = \int_{\phi(S)} d\lambda(s) \int_u^{+\infty} dx \mathbb{E} (\delta^k(Y''(s))/Y(s) = x, Y'(s) = 0) p_{Y(s), Y'(s)}(x, 0). \quad (6)$$

Since $M_{u,k}^X(S)$ is equal to $M_{u,k}^Y\{\phi(S)\}$, we see that the result is a direct consequence of Theorem (2.3)

Even though in the integrand in (5) the product does not depend on the parameterization, each factor does. When the manifold U is equipped with a Riemannian metric it is possible to rewrite equation (5) as

$$\int_u^{+\infty} dx \mathbb{E} (\delta^k(\nabla^2 X(s)/X(s) = x, \nabla X(s) = 0) p_{X(s), \nabla X(s)}(x, 0) Vol \quad (7)$$

where $\nabla^2 X(s)$ and $\nabla X(s)$ are respectively the Hessian and the gradient read in an orthonormal basis.

Remark: One can consider a number of variants of Rice formulae, in which we may be interested in computing the moments of the number of roots of the equation $Z(t) = u$ under some additional conditions. This has been the case in the statement of Theorem 2.3 in which we have given formulae for the first moment of the number of zeroes of X' in which X is bigger than u ($i=2$) and also the real-valued process X has a local maximum ($i=1$).

We just consider below two additional examples of variants that we state here for further reference. We limit the statements to random fields defined on subsets of \mathcal{R}^d . Similar statements hold true when the parameter set is a general smooth manifold. Proofs are essentially the same as the previous ones.

Variante 1: Assume that Z_1, Z_2 are \mathcal{R}^d -valued random fields defined on compact subsets I_1, I_2 of \mathcal{R}^d and suppose that (Z_i, I_i) ($i = 1, 2$) satisfy the hypotheses of Theorem 2.1 and that for every $s \in I_1$ and $t \in I_2$, the distribution of $(Z_1(s), Z_2(t))$ does not degenerate. Then, for each pair $u_1, u_2 \in \mathcal{R}^d$:

$$\mathbb{E} (N_{u_1}^{Z_1}(I_1) N_{u_2}^{Z_2}(I_2)) = \int_{I_1 \times I_2} dt_1 dt_2 \mathbb{E} (|\det(Z_1'(t_1))| |\det(Z_2'(t_2))| / Z_1(t_1) = u_1, Z_2(t_2) = u_2) p_{Z_1(t_1), Z_2(t_2)}(u_1, u_2), \quad (8)$$

Variante 2: Let Z, I be as in Theorem 2.1 and ξ a real-valued bounded random variable which is measurable with respect to the σ -algebra generated by the process Z . Assume that for each $t \in I$, there exists a continuous Gaussian process $\{Y^t(s) : s \in I\}$, for each $s, t \in I$ a non-random function $\alpha^t(s) : \mathcal{R}^d \rightarrow \mathcal{R}^d$ and a Borel-measurable function $g : \mathcal{C} \rightarrow \mathcal{R}$ where \mathcal{C} is space of real-valued continuous functions on I equipped with the supremum norm, such that:

1. $\xi = g(Y^t(\cdot) + \alpha^t(\cdot)Z(t))$
2. $Y^t(\cdot)$ and $Z(t)$ are independent
3. for each $u_0 \in \mathcal{R}$, almost surely the function $u \rightsquigarrow g(Y^t(\cdot) + \alpha^t(\cdot)u)$ is continuous at $u = u_0$

Then the formula :

$$\mathbb{E} (N_u^Z(I)\xi) = \int_I \mathbb{E} (|\det(Z'(t))|\xi/Z(t) = u) p_{Z(t)}(u)dt,$$

holds true.

We will be particularly interested in the function $\xi = \mathbb{1}_{M_I < v}$ for some $v \in \mathcal{R}$. We will see that later on that it satisfies the above conditions under certain hypotheses on the process Z .

3 First Derivative, First Form.

Our main goals in this and the next section are to prove existence and regularity of the derivative of the function $u \rightsquigarrow F_I(u)$ and, at the same time, that it satisfies some implicit formulae that can be used to provide bounds on it. In the following we assume that I is a d -dimensional \mathcal{C}^∞ manifold embedded in \mathcal{R}^N , $N \geq d$. σ and $\tilde{\sigma}$ are respectively the geometric measures on I and ∂I . Unless explicit statement of the contrary, the topology on I will be the relative topology.

In this section we prove formula (10) for $F'_I(u)$. -that we call “first form”- which is valid for λ -almost every u , under strong regularity conditions on the paths of the process X . In fact, the hypothesis that X is Gaussian is only used in Rice formula itself and in Lemma 3.1 which gives a bound for the joint density $p_{X(s), X(t), X'(s), X'(t)}$. In both places, one can substitute Gaussianity by appropriate conditions that permit to obtain similar results.

More generally, it is easy to see that inequality (9) below is valid under quite general non Gaussian conditions and implies that we have the following upper bound for the density of the distribution of the random variable M_I .

$$F'_I(u) \leq \int_I \mathbb{E} (\delta^1(X''(t))/X(t) = u, X'(t) = 0) p_{X(t), X'(t)}(u, 0) \sigma(dt) + \int_{\partial I} \mathbb{E} (\delta^1(\tilde{X}''(t))/X(t) = u, \tilde{X}'(t) = 0) p_{X(t), \tilde{X}'(t)}(u, 0) \tilde{\sigma}(dt), \quad (9)$$

where the function δ^1 has been defined in the statement of Theorem 2.3 and \tilde{X} denotes the restriction of X to the boundary ∂I .

Even for $d = 1$ (one parameter processes) and X Gaussian and stationary, inequality (9) provides reasonably good upper bounds for $F'_I(u)$ (see Diebolt and Posse (1996), Azaïs and Wschebor (2001)). We will see an example for $d = 2$ at the end of this section.

In the next section, we are able to prove that $F_I(u)$ is a \mathcal{C}^1 function and that formula (10) can be essentially simplified by getting rid of the conditional expectation, thus obtaining the “second form” for the derivative. This is done under weaker regularity conditions but the assumption that X is Gaussian becomes essential.

Definition 3.1 *Let $X : I \rightarrow \mathcal{R}$ be real-valued stochastic process defined on a subset of \mathcal{R}^d . We will say that X satisfies condition (H_k) , k a positive integer, if the following three conditions hold true:*

- X is Gaussian;
- a.s. the paths of X are of class C^k ;
- for any choice of pairwise different values of the parameter t_1, \dots, t_n the joint distribution of the random variables $X(t_1), \dots, X(t_n), X'(t_1), \dots, X'(t_n), \dots, X^{(k)}(t_1), \dots, X^{(k)}(t_n)$ has maximum rank.

The next proposition shows that there exist processes that satisfy (H_k) .

Proposition 3.1 *Let $X = \{X(t) : t \in \mathcal{R}^d\}$ be a centred stationary Gaussian process having continuous spectral density f^X . Assume that $f^X(x) > 0$ for every $x \in \mathcal{R}^d$ and that for any $\alpha > 0$ $f^X(x) \leq C_\alpha \|x\|^{-\alpha}$ holds true for some constant C_α and all $x \in \mathcal{R}^d$. Then, X satisfies (H_k) for every $k = 1, 2, \dots$*

Proof: The proof is an adaptation of the proof of a related result for $d = 1$ (Cramer & Leadbetter (1967), p. 203) see Azaïs and Wschebor (2002) \square

Theorem 3.1 (First derivative, first form) *Let $X : I \rightarrow \mathcal{R}$ be a Gaussian process, I a C^∞ compact d -dimensional manifold. Assume that X verifies H_k for every $k = 1, 2, \dots$*

Then, the function $u \rightsquigarrow F_I(u)$ is absolutely continuous and its Radon-Nikodym derivative is given for almost every u by:

$$F'_I(u) = (-1)^d \int_I \mathbf{E}(\det(X''(t)) \mathbb{1}_{M_I \leq u} / X(t) = u, X'(t) = 0) p_{X(t), X'(t)}(u, 0) \sigma(dt) + (-1)^{d-1} \int_{\partial I} \mathbf{E}(\det(\tilde{X}''(t)) \mathbb{1}_{M_I \leq u} / X(t) = u, \tilde{X}'(t) = 0) p_{X(t), \tilde{X}'(t)}(u, 0) \tilde{\sigma}(dt). \quad (10)$$

Proof: For $u < v$ and S (respectively \tilde{S}) a subset of I (resp. ∂I), let us denote

$$\begin{aligned} M_{u,v}(S) &= \#\{t \in S : u < X(t) \leq v, X'(t) = 0, X''(t) \prec 0\} \\ \tilde{M}_{u,v}(\tilde{S}) &= \#\{t \in \tilde{S} : u < X(t) \leq v, \tilde{X}'(t) = 0, \tilde{X}''(t) \prec 0\} \end{aligned}$$

Step 1. Let $h > 0$ and consider the increment

$$F_I(u) - F_I(u - h) = \mathbf{P}\left(\{M_I \leq u\} \cap \left[\{M_{u-h,u}(\dot{I}) \geq 1\} \cup \{\tilde{M}_{u-h,u}(\partial I) \geq 1\}\right]\right).$$

Let us prove that

$$\mathbf{P}\left(M_{u-h,u}(\dot{I}) \geq 1, \tilde{M}_{u-h,u}(\partial I) \geq 1\right) = o(h) \text{ as } h \downarrow 0. \quad (11)$$

In fact, for $\delta > 0$:

$$\begin{aligned} \mathbf{P}\left(M_{u-h,u}(\dot{I}) \geq 1, \tilde{M}_{u-h,u}(\partial I) \geq 1\right) \\ \leq \mathbf{E}\left(M_{u-h,u}(I_{-\delta}) \tilde{M}_{u-h,u}(\partial I)\right) + E(M_{u-h,u}(I \setminus I_{-\delta})) \end{aligned} \quad (12)$$

The first term in the right-hand member of (12) can be computed by means of a Rice-type Formula, and it can be expressed as:

$$\begin{aligned} \int_{I_{-\delta} \times \partial I} \sigma(dt) \tilde{\sigma}(d\tilde{t}) \int \int_{u-h}^u dx d\tilde{x} \\ \mathbf{E}\left(\delta^1(X''(t)) \delta^1(\tilde{X}''(\tilde{t})) / X(t) = x, \tilde{X}(\tilde{t}) = \tilde{x}, X'(t) = 0, \tilde{X}'(\tilde{t}) = 0\right) \\ p_{X(t), \tilde{X}(\tilde{t}), X'(t), \tilde{X}'(\tilde{t})}(x, \tilde{x}, 0, 0), \end{aligned}$$

where the function δ^1 has been defined in Theorem 2.3.

Since in this integral $\|t - \tilde{t}\| \geq \delta$, the integrand is bounded and the integral is $O(h^2)$.

For the second term in (12) we apply Rice formula again. Taking into account that the boundary of I is smooth and compact, we get:

$$\begin{aligned} & \mathbb{E}(M_{u-h,u}(I \setminus I_{-\delta})) \\ &= \int_{I \setminus I_{-\delta}} \sigma(dt) \int_{u-h}^u \mathbb{E}(\delta^1(X''(t))/X(t) = x, X'(t) = 0) p_{X(t), X'(t)}(x, 0) dx \\ & \leq (\text{const}) h \sigma(I \setminus I_{-\delta}) \leq (\text{const}) h\delta, \end{aligned}$$

where the constant does not depend on h and δ . Since $\delta > 0$ can be chosen arbitrarily small, (11) follows and we may write as $h \rightarrow 0$:

$$\begin{aligned} & F_I(u) - F_I(u-h) \\ &= \mathbb{P}\left(M_I \leq u, M_{u-h,u}(\dot{I}) \geq 1\right) + \mathbb{P}\left(M_I \leq u, \widetilde{M}_{u-h,u}(\partial I) \geq 1\right) + o(h) \end{aligned}$$

Note that the foregoing argument also implies that F_I is absolutely continuous with respect to Lebesgue measure and that the density is bounded above by the right-hand member of (10). In fact:

$$\begin{aligned} F_I(u) - F_I(u-h) &\leq \mathbb{P}\left(M_{u-h,u}(\dot{I}) \geq 1\right) + \mathbb{P}\left(\widetilde{M}_{u-h,u}(\partial I) \geq 1\right) \\ &\leq \mathbb{E}\left(M_{u-h,u}(\dot{I})\right) + \mathbb{E}\left(\widetilde{M}_{u-h,u}(\partial I)\right) \end{aligned}$$

and it is enough to apply Rice Formula to each one of the expectations on the right-hand side.

The delicate part of the proof consists in showing that we have equality in (10).

Step 2. For $g : I \rightarrow \mathcal{R}$ we put $\|g\|_\infty = \sup_{t \in I} |g(t)|$ and if k is a non-negative integer, $\|g\|_{\infty, k} = \sup_{k_1+k_2+\dots+k_d \leq k} \|\partial_{k_1, k_2, \dots, k_d} g\|_\infty$. For fixed $\gamma > 0$ (to be chosen later on) and $h > 0$, we denote by $E_h = \left\{ \|X\|_{\infty, 4} \leq h^{-\gamma} \right\}$. Because of the Landau-Shepp-Fernique inequality (see Landau-Shepp, 1970 or Fernique, 1975) there exist positive constants C_1, C_2 such that

$$\mathbb{P}(E_h^C) \leq C_1 \exp[-C_2 h^{-2\gamma}] = o(h) \text{ as } h \rightarrow 0$$

so that to have (10) it suffices to show that, as $h \rightarrow 0$:

$$\mathbb{E} \left(\left[M_{u-h,u}(\dot{I}) - \mathbb{1}_{M_{u-h,u}(\dot{I}) \geq 1} \right] \mathbb{1}_{M_I \leq u} \mathbb{1}_{E_h} \right) = o(h) \quad (13)$$

$$\mathbb{E} \left(\left[\widetilde{M}_{u-h,u}(\partial I) - \mathbb{1}_{\widetilde{M}_{u-h,u}(\partial I) \geq 1} \right] \mathbb{1}_{M_I \leq u} \mathbb{1}_{E_h} \right) = o(h) \quad (14)$$

We prove (13). (14) can be proved in a similar way.

Put $M_{u-h,u} = M_{u-h,u}(\dot{I})$. We have, on applying Rice formula for the second factorial moment:

$$\begin{aligned} \mathbb{E} \left(\left[M_{u-h,u} - \mathbb{1}_{M_{u-h,u} \geq 1} \right] \mathbb{1}_{M_I \leq u} \mathbb{1}_{E_h} \right) &\leq \mathbb{E} (M_{u-h,u} (M_{u-h,u} - 1) \mathbb{1}_{E_h}) \\ &= \iint_{I \times I} A_{s,t} \sigma(ds) \sigma(dt), \end{aligned} \quad (15)$$

where

$$\begin{aligned} A_{s,t} &= \iint_{u-h}^u dx_1 dx_2 \\ \mathbb{E} \left(|\det(X''(s) \det(X''(t))| \mathbb{1}_{X''(s) < 0, X''(t) < 0} \mathbb{1}_{E_h} / X(s) = x_1, X(t) = x_2, X'(s) = 0, X'(t) = 0) \right. \\ &\quad \left. \cdot p_{X(s), X(t), X'(s), X'(t)}(x_1, x_2, 0, 0) \right). \end{aligned} \quad (16)$$

Our goal is to prove that $A_{s,t}$ is $o(h)$ as $h \downarrow 0$ uniformly on s, t . Note that when s, t vary in a domain of the form $D_\delta := \{t, s \in I : \|t - s\| > \delta\}$ for some $\delta > 0$, then the Gaussian distribution in (16) is non-degenerate and $A_{s,t}$ is bounded by $(const)h^2$, the constant depending on the minimum of the determinant: $\det \text{Var}((X(s), X(t), X'(s), X'(t)))$, for $s, t \in D_\delta$.

So it is enough to prove that $A_{s,t} = o(h)$ for $\|t - s\|$ small, and we may assume that s and t are in the same chart (U, ϕ) . Writing the process in this chart we may assume that I is a ball or a half ball in \mathcal{R}^d . Let s, t two such points, define the process $Y = Y^{s,t}$ by $Y(\tau) = X(s + \tau(t - s))$; $\tau \in [0, 1]$. Under the conditioning one has:

$$\begin{aligned} Y(0) &= x_1, \quad Y(1) = x_2, \quad Y'(0) = Y'(1) = 0 \\ Y''(0) &= X''(s)[(t - s), (t - s)] \quad ; \quad Y''(1) = X''(t)[(t - s), (t - s)]. \end{aligned}$$

Consider the interpolation polynomial Q of degree 3 such that

$$Q(0) = x_1, \quad Q(1) = x_2, \quad Q'(0) = Q'(1) = 0$$

Check that

$$Q(y) = x_1 + (x_2 - x_1) y^2(3 - 2y), \quad Q''(0) = -Q''(1) = 6(x_2 - x_1)$$

Denote $Z(\tau) = Y(\tau) - Q(\tau)$ $0 \leq \tau \leq 1$. Under the conditioning, one has $Z(0) = Z(1) = Z'(0) = Z'(1) = 0$ and if also the event E_h occurs, an elementary calculation shows that for $0 \leq \tau \leq 1$:

$$|Z''(\tau)| \leq \sup_{\tau \in [0,1]} \frac{|Z^{(4)}(\tau)|}{2!} = \sup_{\tau \in [0,1]} \frac{|Y^{(4)}(\tau)|}{2!} \leq (\text{const}) \|t - s\|^4 h^{-\gamma}. \quad (17)$$

On the other hand, check that if A is a positive semi-definite symmetric $d \times d$ real matrix and v_1 is a vector of Euclidean norm equal to 1, then the inequality

$$\det(A) \leq \langle Av_1, v_1 \rangle \det(B) \quad (18)$$

holds true, where B is the $(d-1) \times (d-1)$ matrix $B = ((\langle Av_j, v_k \rangle))_{j,k=2,\dots,d}$ and $\{v_1, v_2, \dots, v_d\}$ an orthonormal basis of R^d containing v_1 .

Assume $X''(s)$ is negative definite, and that the event E_h occurs. We can apply (18) to the matrix $A = -X''(s)$ and the unit vector $v_1 = (t-s)/\|t-s\|$. Note that in that case, the elements of matrix B are of the form $\langle -X''(s)v_j, v_k \rangle$ hence bounded by $(\text{const})h^{-\gamma}$. So,

$$\det[-X''(s)] \leq \langle -X''(s)v_1, v_1 \rangle C_d h^{-(d-1)\gamma} = C_d [Y''(0)]^- \|t-s\|^{-2} h^{-(d-1)\gamma}$$

the constant C_d depending only on the dimension d .

Similarly, if $X''(t)$ is negative definite, and the event E_h occurs, then:

$$\det[-X''(t)] \leq C_d [Y''(1)]^- \|t-s\|^{-2} h^{-(d-1)\gamma}$$

Hence, if \mathcal{C} is the condition $\{X(s) = x_1, X(t) = x_2, X'(s) = 0, X'(t) = 0\}$:

$$\begin{aligned} & \mathbb{E} (|\det(X''(s)) \det(X''(t))| \mathbb{1}_{X''(s) < 0, X''(t) < 0} \mathbb{1}_{E_h} / \mathcal{C}) \\ & \leq C_d^2 h^{-2(d-1)\gamma} \|t-s\|^{-4} \mathbb{E} \left([Y''(0)]^- [Y''(1)]^- \mathbb{1}_{E_h} / \mathcal{C} \right) \\ & \leq C_d^2 h^{-2(d-1)\gamma} \|t-s\|^{-4} \mathbb{E} \left(\left[\frac{Y''(0) + Y''(1)}{2} \right]^2 \mathbb{1}_{E_h} / \mathcal{C} \right) \\ & = C_d^2 h^{-2(d-1)\gamma} \|t-s\|^{-4} \mathbb{E} \left(\left[\frac{Z''(0) + Z''(1)}{2} \right]^2 \mathbb{1}_{E_h} / \mathcal{C} \right) \\ & \leq (\text{const}) C_d^2 h^{-2d\gamma} \|t-s\|^4 \end{aligned}$$

We now turn to the density in (15) using the following Lemma which is similar to Lemma 4.3., p. 76, in Piterbarg (1996). The proof is omitted.

Lemma 3.1 For all $s, t \in I$:

$$\|t - s\|^{d+3} p_{X(s), X(t), X'(s), X'(t)}(0, 0, 0, 0) \leq D \quad (19)$$

where D is a constant.

Back to the proof of the theorem, to bound the expression in (15) we use Lemma 3.1 and the bound on the conditional expectation, thus obtaining

$$\begin{aligned} \mathbb{E}(M_{u-h,u}(M_{u-h,u} - 1)\mathbb{1}_{E_h}) &\leq (\text{const})C_d^2 h^{-2d\gamma} D \\ &\iint_{I \times I} \|t - s\|^{-d+1} ds dt \iint_{u-h}^u dx_1 dx_2 \leq (\text{const}) h^{2-2d\gamma} \end{aligned} \quad (20)$$

since the function $(s, t) \rightsquigarrow \|t - s\|^{-d+1}$ is Lebesgue-integrable in $I \times I$. The last constant depends only on the dimension d and the set I , Taking γ small enough (13) follows. \square

An example: Let $\{X(s, t)\}$ be a real-valued two-parameter Gaussian, centred stationary isotropic process with covariance Γ . Assume that $\Gamma(0) = 1$ and that the spectral measure μ is absolutely continuous with density $\mu(ds, dt) = f(\rho)dsdt$, $\rho = (s^2 + t^2)^{\frac{1}{2}}$. Assume further that $J_k = \int_0^{+\infty} \rho^k f(\rho)d\rho < \infty$, for $1 \leq k \leq 5$. Our aim is to give an explicit upper bound for the density of the probability distribution of M_I where I is the unit disc. Using (9) which is a consequence of Theorem 3.1 and the invariance of the law of the process, we have

$$\begin{aligned} F'_I(u) &\leq \pi \mathbb{E}\left(\delta^1(X''(0, 0))/X(0, 0) = u, X'(0, 0) = (0, 0)\right) p_{X(0,0), X'(0,0)}(u, (0, 0)) \\ &+ 2\pi \mathbb{E}\left(\delta^1(\tilde{X}''(1, 0))/X(1, 0) = u, \tilde{X}'(1, 0) = 0\right) p_{X(1,0), \tilde{X}'(1,0)}(u, 0) = I_1 + I_2. \end{aligned} \quad (21)$$

We denote by X, X', X'' the value of the different processes at some point (s, t) ; by $X''_{ss}, X''_{st}, X''_{tt}$ the entries of the matrix X'' and by φ and Φ the standard normal density and distribution.

One can easily check that: X' is independent of X and X'' , and has variance $\pi J_3 I_d$; X''_{st} is independent of X, X', X''_{ss} and X''_{tt} , and has variance $\frac{\pi}{4} J_5$. Conditionally on $X = u$, the random variables X''_{ss} and X''_{tt} have expectation: $-\pi J_3$; variance: $\frac{3\pi}{4} J_5 - (\pi J_3)^2$; covariance: $\frac{\pi}{4} J_5 - (\pi J_3)^2$.

We obtain

$$I_2 = \sqrt{\frac{2}{J_3}} \varphi(u) \left[\left(\frac{3\pi}{4} J_5 - (\pi J_3)^2\right)^{\frac{1}{2}} \varphi(bu) + \pi J_3 u \Phi(bu) \right],$$

with $b = \frac{\pi J_3}{\left(\frac{3\pi}{4} J_5 - (\pi J_3)^2\right)^{\frac{1}{2}}}$.

As for I_1 we remark that, conditionally on $X = u$, $X''_{ss} + X''_{tt}$ and $X''_{ss} - X''_{tt}$ are independent, so that a direct computation gives:

$$I_1 = \frac{1}{8\pi J_3} \varphi(u) \mathbb{E} \left[(\alpha \eta_1 - 2\pi J_3 u)^2 - \frac{\pi J_5}{4} (\eta_2^2 + \eta_3^2) \right. \\ \left. \mathbb{1}_{\{\alpha \eta_1 < 2\pi J_3 u\}} \mathbb{1}_{\{(\alpha \eta_1 - 2\pi J_3 u)^2 - \frac{\pi J_5}{4} (\eta_2^2 + \eta_3^2) > 0\}} \right], \quad (22)$$

Where η_1, η_2, η_3 are standard independent normal random variables and $\alpha^2 = 2\pi J_5 - 4\pi^2 J_3^2$. Finally we get

$$I_1 = \frac{\sqrt{2\pi}}{8\pi J_3} \varphi(u) \int_0^\infty \left[(\alpha^2 + a^2 - c^2 x^2) \Phi(a - cx) + [2a\alpha - \alpha^2(a - cx)] \varphi(a - cx) \right] x \varphi(x) dx,$$

with $a = 2\pi J_3 u$, $c = \sqrt{\frac{\pi J_5}{4}}$.

4 First derivative, second form

We choose, once for all along this section a finite atlas \mathcal{A} for I . Then, to every $t \in I$ it is possible to associate a fixed chart that will be denoted (U_t, ϕ_t) . When $t \in \partial I$, $\phi_t(U_t)$ can be chosen to be a half ball with $\phi_t(t)$ belonging to the hyperplane limiting this half ball. For $t \in I$, let V_t an open neighbourhood of t whose closure is included in U_t and ψ_t a C^∞ function such that $\psi_t \equiv 1$ on V_t ; $\psi_t \equiv 0$ on U_t^c .

- For every $t \in \dot{I}$ and $s \in I$ we define the normalization $n(t, s)$ in the following way:

- for $s \in V_t$, we set “in the chart” (U_t, ϕ_t) $n_1(t, s) = \frac{1}{2} \|s - t\|^2$.
By “in the chart” we mean that $\|s - t\|$, is in fact $\|\phi_t(t) - \phi_t(s)\|$.
- for general s we set $n(t, s) = \psi_t(s) n_1(t, s) + (1 - \psi_t(s))$

Note that in the flat case ($d=N$) the simpler definition $n(t, s) = \frac{1}{2} \|s - t\|^2$ works.

- For every $t \in \partial I$ and $s \in I$, we set $n_1(t, s) = |(s - t)_N| + \frac{1}{2} \|s - t\|^2$, where $(s - t)_N$ is the normal component of $(s - t)$ with respect to the hyperplane delimiting the half ball $\phi_t(U_t)$. The rest of the definition is the same.

Definition 4.1 We will say that f is an helix-function - or an h-function - on I with pole $t \in I$ satisfying hypothesis $H_{t,k}$, k integer $k > 1$ if

- f is a bounded \mathcal{C}^k function on $I \setminus \{t\}$.
- $\underline{f}(s) := n(t, s)f(s)$ can be prolonged as function of class \mathcal{C}^k on I .

Definition 4.2 In the same way X is called an h-process with pole $t \in I$ satisfying hypothesis $H_{t,k}$, k integer $k > 1$ if

- Z is a Gaussian process with \mathcal{C}^k paths on $I \setminus \{t\}$.
- for $t \in \dot{I}$; $\underline{Z}(s) := n(t, s)Z(s)$ can be prolonged as a process of class \mathcal{C}^k on I , with $\underline{Z}(t) = 0$ $\underline{Z}'(t) = 0$. If s_1, \dots, s_m are pairwise different points of $I \setminus \{t\}$ then the distribution of $\underline{Z}^{(2)}(t), \dots, \underline{Z}^{(k)}(t), \underline{Z}(s_1), \dots, \underline{Z}^{(k)}(s_1), \dots, \underline{Z}^{(k)}(s_m)$ does not degenerate.
- for $t \in \partial I$; $\underline{Z}(s) := n(t, s)Z(s)$ can be prolonged as a process of class \mathcal{C}^k on I with $\underline{Z}(t) = 0$ $\underline{Z}'(t) = 0$ and if s_1, \dots, s_m are pairwise different points of $I \setminus \{t\}$ then the distribution of $\underline{Z}'_N(t), \underline{Z}^{(2)}(t), \dots, \underline{Z}^{(k)}(t), \underline{Z}(s_1), \dots, \underline{Z}^{(k)}(s_1), \dots, \underline{Z}^{(k)}(s_m)$ does not degenerate. $\underline{Z}'_N(t)$ is the derivative normal to the boundary of I at t .

We use the terms “h-function” and “h-process” since the function and the paths of the process need not to extend to a continuous function at the point t . However, the definition implies the existence of radial limits at t . So the process may take the form of a helix around t .

Lemma 4.1 Let X be a process satisfying H_k , $k \geq 2$, and f be a \mathcal{C}^k function $I \rightarrow \mathcal{R}$
(A) For $t \in \dot{I}$, set for $s \in I, s \neq t$

$$X(s) = a_s^t X(t) + \langle b_s^t, X'(t) \rangle + n(t, s)X^t(s),$$

where a_s^t and b_s^t are the regression coefficients.

In the same way, set

$$f(s) = a_s^t f(t) + \langle b_s^t, f'(t) \rangle + n(t, s)f^t(s),$$

using the regression coefficients associated to X .

(B) For $t \in \partial I$, $s \in T, s \neq t$ set

$$X(s) = \tilde{a}_s^t X(t) + \langle \tilde{b}_s^t, \tilde{X}'(t) \rangle + n(t, s)X^t(s)$$

and

$$f(s) = \tilde{a}_s^t f(t) + \langle \tilde{b}_s^t, \tilde{f}'(t) \rangle + n(t, s) f^t(s),$$

Then $s \rightsquigarrow X^t(s)$ and $s \rightsquigarrow f^t(s)$ are respectively a h -process and a h -function with pole t satisfying $H_{t,k}$.

Proof: We give the proof in the case $t \in \dot{I}$, the other one being similar. In fact, the quantity denoted by $\underline{X}^t(s)$ is just $X(s) - a_s^t X(t) - \langle b_s^t, X'(t) \rangle$. On $L^2(\Omega, P)$, let Π be the projector on the orthogonal complement to the subspace generated by $X(t), X'(t)$. Using a Taylor expansion

$$X(s) = X(t) + \langle (s-t), X'(t) \rangle + \|t-s\|^2 \int_0^1 X''((1-\alpha)t + \alpha s) [v, v] (1-\alpha) d\alpha,$$

With $v = \frac{s-t}{\|s-t\|}$. This implies that

$$\underline{X}^t(s) = \Pi \left[\|t-s\|^2 \int_0^1 X''((1-\alpha)t + \alpha s) [v, v] (1-\alpha) d\alpha \right], \quad (23)$$

which gives the result due to the non degeneracy condition. \square

We state now an extension of Ylvisaker's Theorem (1968) on the regularity of the distribution of the maximum of a Gaussian process which we will use in the proof of Theorem 4.2 and might have some interest in itself.

Theorem 4.1 *Let $Z : T \rightarrow \mathcal{R}$ a Gaussian separable process on some parameter set T and denote by $M^Z = \sup_{t \in T} Z(t)$ which is a random variable taking values in $\mathcal{R} \cup \{+\infty\}$. Assume that there exists $\sigma_0 > 0$, $m_- > -\infty$ such that*

$$m(t) = \mathbb{E}(Z_t) \geq m_- \quad ; \quad \sigma^2(t) = \text{Var}(Z_t) \geq \sigma_0^2$$

for every $t \in T$. Then the distribution of the random variable M^Z is the sum of an atom at $+\infty$ and a possibly defective-probability measure on \mathcal{R} which has a locally bounded density.

Proof: Suppose first that $X : T \rightarrow \mathcal{R}$ a Gaussian separable process satisfying $\text{Var}(X_t) = 1$; $\mathbb{E}(X_t) \geq 0$, for every $t \in T$. A close look at Ylvisaker's proof (1968) shows that the distribution of the supremum M^X has a density p_{M^X} that satisfies

$$p_{M^X}(u) \leq \psi(u) = \frac{\exp(-u^2/2)}{\int_u^\infty \exp(-v^2/2) dv} \text{ for every } u \in \mathcal{R} \quad (24)$$

Let now Z satisfy the hypotheses of the theorem. For given $a, b \in \mathcal{R}, a < b$, choose $A \in \mathcal{R}^+$ so that $|a| < A$ and consider the process:

$$X(t) = \frac{Z(t) - a}{\sigma(t)} + \frac{|m_-| + A}{\sigma_0}.$$

Clearly for every $t \in T$:

$$\mathbb{E}(X(t)) = \frac{m(t) - a}{\sigma(t)} + \frac{|m_-| + A}{\sigma_0} \geq -\frac{|m_-| + |a|}{\sigma_0} + \frac{|m_-| + A}{\sigma_0} \geq 0,$$

and $\text{Var}(X(t)) = 1$. So that (24) holds for the process X .

On the other hand the statement follows from the inclusion:

$$\{a < M^Z \leq b\} \subset \left\{ \frac{|m_-| + A}{\sigma_0} < M^X \leq \frac{|m_-| + A}{\sigma_0} + \frac{b - a}{\sigma_0} \right\}.$$

which implies

$$\mathbb{P}\{a < M^Z \leq b\} \leq \int_{\frac{|m_-| + A}{\sigma_0}}^{\frac{|m_-| + A}{\sigma_0} + \frac{b - a}{\sigma_0}} \psi(u) du = \int_a^b \frac{1}{\sigma_0} \psi\left(\frac{v - a + |m_-| + A}{\sigma_0}\right) dv. \square$$

Set now $\beta(t) \equiv 1$. The key point is that, due to regression formulae, under the condition $\{X(t) = u, X'(t) = 0\}$ the event

$$A_u(X, \beta) := \{X(s) \leq u, \forall s \in I\}$$

coincides with the event

$$A_u(X^t, \beta^t) := \{X^t(s) \leq \beta^t(s)u, \forall s \in I \setminus \{t\}\},$$

where X^t and β^t are the h-process and the h-function defined in Lemma 4.1.

Theorem 4.2 (First derivative, second form) *Let $X : I \rightarrow \mathcal{R}$ be a Gaussian process, I a C^∞ compact manifold contained in \mathcal{R}^d . Assume that X has paths of class C^2 and for $s \neq t$ the triplet $(X(s), X(t), X'(t))$ in $\mathcal{R} \times \mathcal{R} \times \mathcal{R}^d$ has a non-degenerate distribution. Then, the result of Theorem 3.1 is valid, the derivative $F'_I(u)$ given by relation (10) can be written as*

$$\begin{aligned} F'_I(u) &= (-1)^d \int_I \mathbb{E} \left[\det(\underline{X}'''(t) - \underline{\beta}'''(t)u) \mathbb{1}_{A_u(X^t, \beta^t)} \right] p_{X(t), X'(t)}(u, 0) \sigma(dt) \\ &\quad + (-1)^{d-1} \int_{\partial I} \mathbb{E} \left[\det(\tilde{X}'''(t) - \tilde{\beta}'''(t)u) \mathbb{1}_{A_u(X^t, \beta^t)} \right] p_{X(t), \tilde{X}'(t)}(u, 0) \tilde{\sigma}(dt), \end{aligned} \quad (25)$$

and this expression is continuous as a function of u .

The notation $\tilde{X}'''(t)$ should be understood in the sense that we first define \underline{X}^t and then calculate its second derivative along ∂I .

Proof: As a first step, assume that the process X satisfies the hypotheses of theorem 3.1, which are stronger than those in the present theorem.

We prove that the first term in (10) can be rewritten as the first term in (25). One can proceed in a similar way with the second term, mutatis mutandis. For that purpose, use the remark just before the statement of Theorem 4.2 and the fact that under the condition $\{X(t) = u, X'(t) = 0\}$, $X''(t)$ is equal to $\underline{X}'''(t) - \underline{\beta}'''(t)u$.

Replacing in the conditional expectation in (10) and on account of the Gaussianity of the process, we get rid of the conditioning and obtain the first term in (25). We now study the continuity of $u \rightsquigarrow F'_I(u)$. The variable u appears at three locations

- in the density $p_{X(t), X'(t)}(u, 0)$ which is clearly continuous
- in

$$\mathbb{E} \left[\det \left(\underline{X}'''(t) - \underline{\beta}'''(t)u \right) \mathbb{1}_{A_u(X^t, \beta^t)} \right]$$

where it occurs twice: in the first factor and in the indicator function.

Due to the integrability of the supremum of bounded Gaussian processes, it is easy to prove that this expression is continuous as a function of the first u .

As for the u in the indicator function, set

$$\xi_v := \det \left(\underline{X}'''(t) - \underline{\beta}'''(t)v \right) \quad (26)$$

and, for $h > 0$, consider the quantity $\mathbb{E} \left[\xi_v \mathbb{1}_{A_u(X^t, \beta^t)} \right] - \mathbb{E} \left[\xi_v \mathbb{1}_{A_{u-h}(X^t, \beta^t)} \right]$ which is equal to

$$\mathbb{E} \left[\xi_v \mathbb{1}_{A_u(X^t, \beta^t) \setminus A_{u-h}(X^t, \beta^t)} \right] - \mathbb{E} \left[\xi_v \mathbb{1}_{A_{u-h}(X^t, \beta^t) \setminus A_u(X^t, \beta^t)} \right] \quad (27)$$

Apply Schwarz's inequality to the first term in (27).

$$\mathbb{E} \left[\xi_v \mathbb{1}_{A_u(X^t, \beta^t) \setminus A_{u-h}(X^t, \beta^t)} \right] \leq \left[\mathbb{E}(\xi_v^2) \mathbb{P}\{A_u(X^t, \beta^t) \setminus A_{u-h}(X^t, \beta^t)\} \right]^{1/2}$$

The event $A_u(X^t, \beta^t) \setminus A_{u-h}(X^t, \beta^t)$ can be described as

$$\forall s \in I \setminus \{t\} : X^t(s) - \beta^t(s)u \leq 0 ; \exists s_0 \in I \setminus \{t\} : X^t(s_0) - \beta^t(s_0)(u - h) > 0$$

This implies that $\beta^t(s_0) > 0$ and that $-\|\beta^t\|_\infty h \leq \sup_{s \in I \setminus \{t\}} X^t(s) - \beta^t(s)u \leq 0$. Now, observe that our improved version of Ylvisaker's theorem (Theorem 4.1), applies to

the process $s \rightsquigarrow X^t(s) - \beta^t(s)u$ defined on $I \setminus \{t\}$. This implies that the first term in (27) tends to zero as $h \downarrow 0$. An analogous argument applies to the second term. Finally, the continuity of $F'_I(u)$ follows from the fact that one can pass to the limit under the integral sign in (25).

To finish the proof we still have to show that the added hypotheses are in fact unnecessary for the validity of the conclusion. Suppose now that the process X satisfies only the hypotheses of the theorem and define

$$X^\epsilon(t) = Z_\epsilon(t) + \epsilon Y(t) \quad (28)$$

where for each $\epsilon > 0$, Z_ϵ is a real-valued Gaussian process defined on I , measurable with respect to the σ -algebra generated by $\{X(t) : t \in I\}$, possessing \mathcal{C}^∞ paths and such that almost surely $Z_\epsilon(t)$, $Z'_\epsilon(t)$, $Z''_\epsilon(t)$ converge uniformly on I to $X(t)$, $X'(t)$, $X''(t)$ respectively as $\epsilon \downarrow 0$. One standard form to construct such an approximation process Z_ϵ is to use a \mathcal{C}^∞ partition of the unity on I and to approximate locally the composition of a chart with the function X by means of a convolution with a \mathcal{C}^∞ kernel.

In (28), Y denotes the restriction to I of a Gaussian centred stationary process satisfying the hypotheses of proposition 3.1, defined on \mathcal{R}^N , and independent of X . Clearly X^ϵ satisfies condition (H_k) for every k , since it has \mathcal{C}^∞ paths and the independence of both terms in (28) ensures that X^ϵ inherits from Y the non-degeneracy condition in Definition 3.1. So, if

$M_I^\epsilon = \max_{t \in I} X^\epsilon(t)$ and $F_I^\epsilon(u) = \mathbb{P}\{M_I^\epsilon \leq u\}$ one has

$$\begin{aligned} F_I^{\epsilon'}(u) &= (-1)^d \int_I \mathbb{E} \left[\det \left(\underline{X}^{\epsilon t''}(t) - \underline{\beta}^{\epsilon t''}(t)u \right) \mathbb{1}_{A_u(X^{\epsilon t}, \beta^{\epsilon t})} \right] p_{X^{\epsilon(t)}, X^{\epsilon'(t)}}(u, 0) \sigma(dt) \\ &+ (-1)^{d-1} \int_{\partial I} \mathbb{E} \left[\det \left(\tilde{\underline{X}}^{\epsilon t''}(t) - \tilde{\underline{\beta}}^{\epsilon t''}(t)u \right) \mathbb{1}_{A_u(X^{\epsilon t}, \beta^{\epsilon t})} \right] p_{X^{\epsilon(t)}, \tilde{X}^{\epsilon'(t)}}(u, 0) \tilde{\sigma}(dt), \quad (29) \end{aligned}$$

We want to pass to the limit as $\epsilon \downarrow 0$ in (29). We prove that the right-hand member is bounded if ϵ is small enough and converges to a continuous function of u as $\epsilon \downarrow 0$. Since $M_I^\epsilon \rightarrow M_I$, this implies that the limit is continuous and coincides with $F'_I(u)$ by a standard argument on convergence of densities. We consider only the first term in (29), the second is similar.

The convergence of X_ϵ and its first and second derivative, together with the non-degeneracy hypothesis imply that uniformly on $t \in I$, as $\epsilon \downarrow 0$

$p_{X^{\epsilon(t)}, X^{\epsilon'(t)}}(u, 0) \rightarrow p_{X(t), X'(t)}(u, 0)$. The same kind of argument can be used for $\det \left(\underline{X}^{\epsilon t''}(t) - \underline{\beta}^{\epsilon t''}(t)u \right)$, on account of the form of the regression coefficients and the

definitions of \underline{X}^t and $\underline{\beta}^t$. The only difficulty is to prove that, for fixed u :

$$\mathbb{P}\{C_\epsilon \Delta C\} \rightarrow 0 \text{ as } \epsilon \downarrow 0, \quad (30)$$

where $C_\epsilon = A_u(X^{\epsilon t}, \beta^{\epsilon t})$ $C = A_u(X^t, \beta^t)$

We prove that

$$\text{a. s. } \mathbb{1}_{C_\epsilon} \rightarrow \mathbb{1}_C \text{ as } \epsilon \downarrow 0, \quad (31)$$

which implies (30). First of all, note that the event

$L = \left\{ \sup_{s \in I \setminus \{t\}} (X^t(s) - \beta^t(s)u) = 0 \right\}$ has zero probability, as already mentioned. Second, from the definition of $X^t(s)$ and the hypothesis, it follows that, as $\epsilon \downarrow 0$, $X^{\epsilon t}(s), \beta^{\epsilon t}(s)$ converge to $X^t(s), \beta^t(s)$ uniformly on $I \setminus \{t\}$. Now, if $\omega \notin C$, there exists $\bar{s} = \bar{s}(\omega) \in I \setminus \{t\}$ such that $X^t(\bar{s}) - \beta^t(\bar{s})u > 0$ and for $\epsilon > 0$ small enough, one has $X^{\epsilon t}(\bar{s}) - \beta^{\epsilon t}(\bar{s})u > 0$, which implies that $\omega \notin C_\epsilon$.

On the other hand, let $\omega \in C \setminus L$. This implies that

$$\sup_{s \in I \setminus \{t\}} (X^t(s) - \beta^t(s)u) < 0.$$

From the above mentioned uniform convergence, it follows that if $\epsilon > 0$ is small enough, then $\sup_{s \in I \setminus \{t\}} (X^{\epsilon t}(s) - \beta^{\epsilon t}(s)u) < 0$, hence $\omega \in C_\epsilon$. (31) follows.

So, we have proved that the limit as $\epsilon \downarrow 0$ of the first term in (29) is equal to the first term in (25).

It remains only to prove that the first term in (25) is a continuous function of u . For this purpose, it suffices to show that the function $u \rightsquigarrow \mathbb{P}\{A_u(X^t, \beta^t)\}$ is continuous. This is a consequence of the inequality

$$\left| \mathbb{P}\{A_{u+h}(X^t, \beta^t)\} - \mathbb{P}\{A_u(X^t, \beta^t)\} \right| \leq \mathbb{P}\left\{ \left| \sup_{s \in I \setminus \{t\}} (X^t(s) - \beta^t(s)u) \right| \leq |h| \sup_{s \in I \setminus \{t\}} |\beta^t(s)| \right\}$$

and of Theorem 4.1, applied once again to the process $s \rightsquigarrow X^t(s) - \beta^t(s)u$ defined on $I \setminus \{t\}$.

5 Asymptotic expansion of $F'(u)$ for large u

Corollary 5.1 *Suppose that the process X satisfies the conditions of Theorem 4.2 and that in addition $\mathbb{E}(X_t) = 0$ and $\text{Var}(X_t) = 1$.*

Then, as $u \rightarrow +\infty$ $F'(u)$ is equivalent to

$$\frac{u^d}{(2\pi)^{(d+1)/2}} e^{-u^2/2} \int_I (\det(\Lambda(t)))^{1/2} dt, \quad (32)$$

where $\Lambda(t)$ is the variance-covariance matrix of $X'(t)$.

Proof: Set $r(s, t) := E(X(s), X(t))$, and for $i, j = 1, d$,

$$r_{i;}(s, t) := \frac{\partial}{\partial s_i} r(s, t); \quad r_{ij;}(s, t) := \frac{\partial^2}{\partial s_i \partial s_j} r(s, t); \quad r_{i;j}(s, t) := \frac{\partial^2}{\partial s_i \partial t_j} r(s, t).$$

For every t, i and j $r_{i;}(t, t) = 0$, $\Lambda_{ij}(t) = r_{i;j}(t, t) = -r_{i;j}(t, t)$. Thus $X(t)$ and $X'(t)$ are independent. Regression formulae imply that $a_s^t = r(s, t)$, $\beta^t(s) = \frac{1-r(t,s)}{n(s,t)}$. This implies that $\underline{\beta}^t(t) = \Lambda(t)$ and that the possible limits values of $\beta^t(s)$ as $s \rightarrow t$ are in the set $\{v^T \Lambda(t)v : v \in S^{d-1}\}$. Due to the non-degeneracy condition these quantities are minorized by a positive constant. On the other hand for $s \neq t$ $\beta^t(s) > 0$. This shows that for every $t \in I$ one has $\inf_{s \in I} \beta^t(s) > 0$. Since for every $t \in I$ the process X^t is bounded it follows that *a.s.* $\mathbb{1}_{A_u(X^t, \beta^t)} \rightarrow 1$ as $u \rightarrow +\infty$. Also

$$\det(\underline{X}^{t''}(t) - \underline{\beta}^{t''}(t)u) \simeq (-1)^d \det(\Lambda(t))u^d.$$

Dominated convergence shows that the first term in (25) is equivalent to

$$\int_I u^d \det(\Lambda^t) (2\pi)^{-1/2} e^{-u^2/2} (2\pi)^{-d/2} (\det(\Lambda^t))^{-1/2} dt = \frac{u^d}{(2\pi)^{(d+1)/2}} e^{-u^2/2} \int_I (\det(\Lambda^t))^{1/2} dt.$$

The same kind of argument shows that the second term is $O(u^{d-1}e^{-u^2/2})$ which completes the proof. \square

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