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# THE MEAN NUMBER OF MAXIMA ABOVE HIGH LEVELS IN GAUSSIAN RANDOM FIELDS

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#### **Abstract**

An asymptotic formula for the mean number of maxima above a level of an n-dimensional stationary Gaussian field has been given by Nosko without proof. In this note a short general proof of this formula is given.

RANDOM FIELDS; MAXIMA ABOVE A LEVEL; GAUSSIAN PROCESSES

#### Introduction

Belyayev [1] attributes to Nosko [2] the following asymptotic result for  $M_{\xi}(S)$ , the mean number of maxima above a level  $\xi$  of an *n*-dimensional stationary Gaussian field X(t),  $t = t_1, \dots, t_n$ , with mean zero, in the Lebesgue measurable set S of measure w(S):

(1) 
$$M_{\xi}(S)/w(S) = (2\pi)^{-\frac{1}{2}(n+1)}\mu^{-n+\frac{1}{2}}|A|^{\frac{1}{2}}\xi^{n-1}\exp(-\xi^2/2\mu)[1+O(1/\xi)],\cdots$$

where |A| is the determinant of the covariance matrix of the partial derivatives  $\partial X/\partial t_i$  and  $\mu$  is the variance of X.

However, a perusal of Nosko's paper reveals that the result is stated without proof. A thorough literature search carried out by the author and some colleagues failed to produce a published proof of the general result. Only a partial proof is given in Nosko [3], applicable only to a two-dimensional isotropic Gaussian field, and the method used cannot be generalized to more than two dimensions.

In this note a short proof of (1) for the general case is given.

#### The exact mean number of maxima above a level

The following exact result is given by Belyayev [1]. Let X(t),  $(t = t_1, \dots, t_n)$  be a separable, stationary Gaussian field with mean zero, twice differentiable in quadratic mean. Let  $X_i(t) = \partial X(t)/\partial t_i$ ,  $X_{ij}(t) = \partial^2 X(t)/\partial t_i \partial t_j$ .

Let the following conditions be satisfied:

$$\max_{i,j} E |X_{ij}(t+h) - X_{ij}(t)|^2 \leq \frac{C}{|\log|h||^{1+\varepsilon}}$$
 for  $C > 0$ ,  $\varepsilon > 0$ ;  $i, j = 1, \dots, n$ .

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378 A. M. HASOFER

We introduce the following notation. Let  $\Delta = (z_{ij})$  be a symmetric  $n \times n$  matrix. Then  $z = \text{Vec}(z_{ij})$  will denote the column vector of length n(n+1)/2 obtained from the symmetric matrix  $(z_{ij})$  by placing the successive columns on and above the main diagonal under one another. Let  $y = (y_i)$  be a vector of length n. Let  $\varphi(x, y, z)$  be the joint density function of X(t),  $X_i(t)$ ,  $X_{ij}(t)$ . Let S be a Lebesgue measurable set in  $\mathbb{R}^n$ , and  $M_{\varepsilon}(S)$  the mean number of maxima of X(t) above  $\xi$  in S. Let w(S) be the Lebesgue measure of S. Let  $|\Delta|$  represent the determinant of  $\Delta$ . Then

(2) 
$$m_{\xi} = M_{\xi}(S)/w(S) = (-1)^n \int_{L} |\Delta| \varphi(x, \mathbf{0}, z) dx dz, \cdots$$

where L is the set of points (x, z) for which  $\Delta$  is negative definite, and  $x > \xi$ .

## Proof of the asymptotic formula

The basic idea of the proof is simply that as  $\xi \to \infty$ , the z-set for which  $\Delta$  is negative definite extends to the whole z-space.

To prove this we analyse more closely the function  $\varphi$ . First we carry out a number of simplifying transformations.

- (1) It is easy to see that there always exists an orthogonal transformation of the field coordinates t which will make the derivatives  $X_i$  uncorrelated.
- (2) We change the scale of the  $t_i$  as well as the scale of X to make the variances of X and the  $X_i$  equal to unity.

By considering the spectral representation of X, it is then trivial to reach the following conclusions

- (1) X and  $X_i$  are uncorrelated for all i;
- (2) X and  $X_{ij}$  are uncorrelated for  $i \neq j$ ;
- (3)  $X_i$  and  $X_j$  are uncorrelated for  $i \neq j$ ;
- (4)  $X_i$  and  $X_{jk}$  are uncorrelated for all i, j, k.

Thus we can write

$$\varphi(\mathbf{x}, \mathbf{y}, \mathbf{z}) = \varphi_1(\mathbf{x})\varphi_2(\mathbf{y})\varphi_3(\mathbf{z} \mid \mathbf{x}),$$

and we note that

$$\varphi_2(\mathbf{0}) = (2\pi)^{-\frac{1}{2}n} = (-1)^n \alpha$$
, say.

Thus

$$m_{\xi} = \alpha \int_{\xi}^{\infty} \varphi_{1}(x) dx \int_{P} |\Delta| \varphi_{3}(z \mid x) dz,$$

where P is the set of z for which  $\Delta$  is negative definite.

On account of (2) above, the only effect of conditioning the variables  $X_{ij}$  on

X = x will be to shift the mean of the  $X_{ij}$  by an amount -x, and to change the covariance matrix in a way independent of x. If we make the change of variables  $z_{ij} = v_{ij} - x\delta_{ij}$ , where  $\delta_{ij}$  is the Kronecker delta, we find

$$m_{\xi} = \alpha \int_{\xi}^{\infty} \varphi_{1}(x) dx \int_{U(x)} |V - xI| \varphi_{4}(v) dv,$$

where  $V = (v_{ij})$ ,  $v = \text{Vec}(v_{ij})$ ,  $\varphi_4(v)$  is a multivariate normal distribution independent of x, and U(x) is the set of points v for which the matrix V - xI is negative definite.

We now claim that by making x sufficiently large, we can make U(x) contain the sphere  $\sigma(R)$  centred at the origin with arbitrary radius R. In fact, since V is symmetric, we can diagonalize it by means of an orthogonal matrix P. Thus

$$P'(V-xI)P = P'VP - xI = \operatorname{diag}(w_1 - x, \dots, w_n - x),$$

say.

Thus  $\Delta$  will be negative definite provided  $x > \max_i w_i$ . But  $\max_i w_i \le ||V||$ , where  $||\cdot||$  denotes any matrix norm. Thus, in particular

$$\max_{i} w_{i} \leq \max_{j} \sum_{i} |v_{ij}| \leq n \max_{i,j} |v_{ij}| \leq n |v|,$$

and U(x) will contain the sphere  $\sigma(R)$  provided x > nR.

Next we note that |V-xI| is the characteristic polynomial of the matrix V, so that the coefficient of  $x^n$  is  $(-1)^n$ . It is now easy to estimate that the relative error made in replacing U(x) by the whole space of v and the determinant  $|\Delta|$  by  $(-1)^n x^n$  is  $O(1/\xi)$ . Thus

$$m_{\xi} = \alpha \int_{\xi}^{\infty} (-1)^n x^n \varphi_1(x) dx [1 + O(1/\xi)].$$

And, since  $\int_{\xi}^{\infty} x^n \varphi_1(x) dx = \xi^{n-1} \varphi_1(\xi) [1 + O(1/\xi)]$ , we obtain the asymptotic formula

$$m_{\xi} = (2\pi)^{-\frac{1}{2}(n+1)} \xi^{n-1} \exp(-\xi^2/2) [1 + O(1/\xi)].$$

Finally, reverting to the original coordinates of t and X, we obtain Nosko's formula (1).

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